

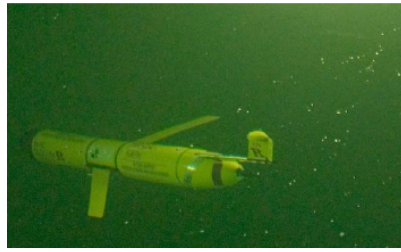
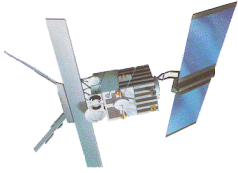
Optimal Spectral Decomposition (OSD) for Ocean Data Analysis

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GTSP Meeting, Honolulu, Hawaii, 27 October 2008



How can we effectively use observational ocean data to represent and to predict the ocean state



Collaborators

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- Chenwu Fan (NPS)
- Tateana Margolina (NPS)
- Oleg Melnichenko (Univ of Hawaii)

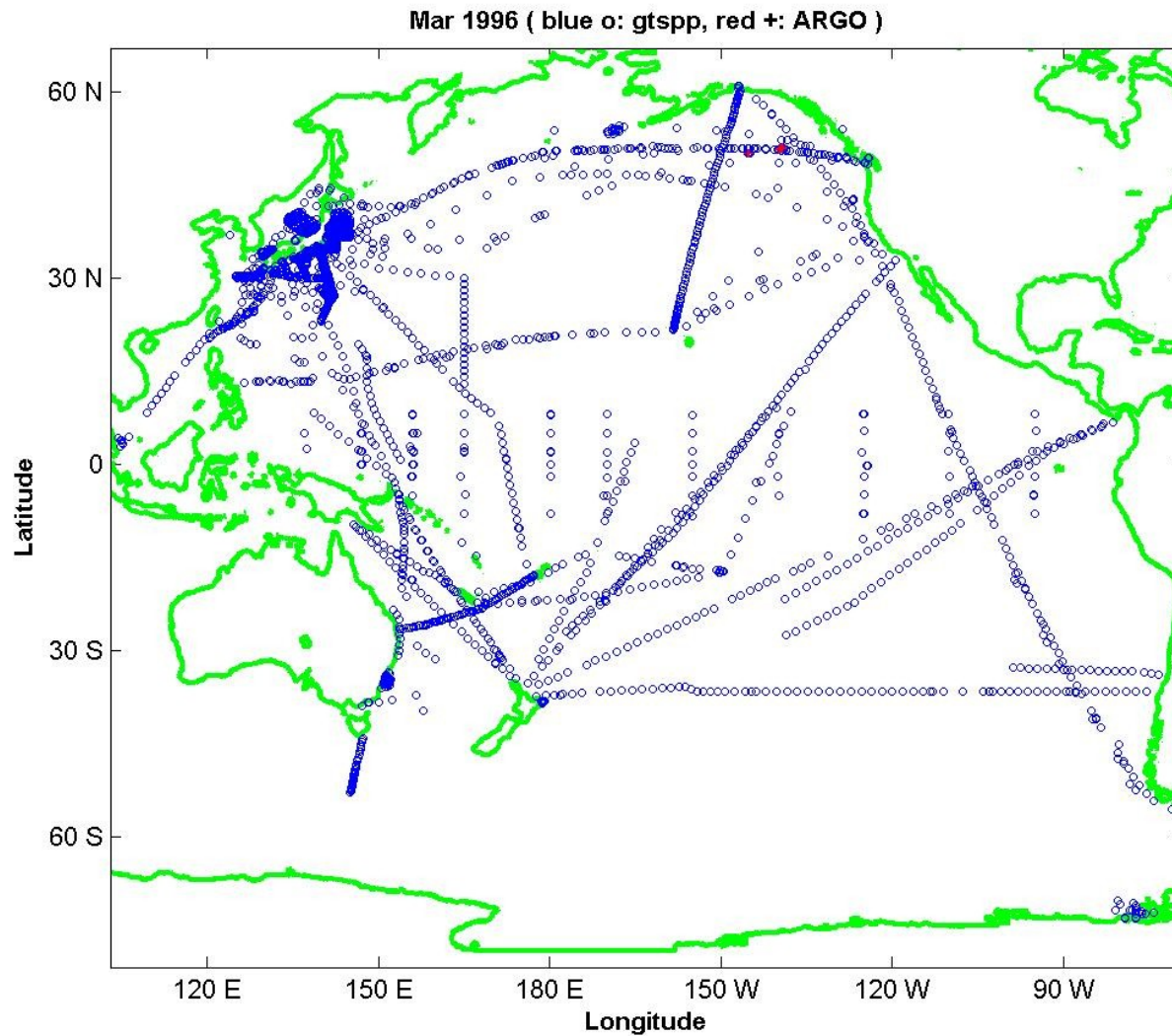
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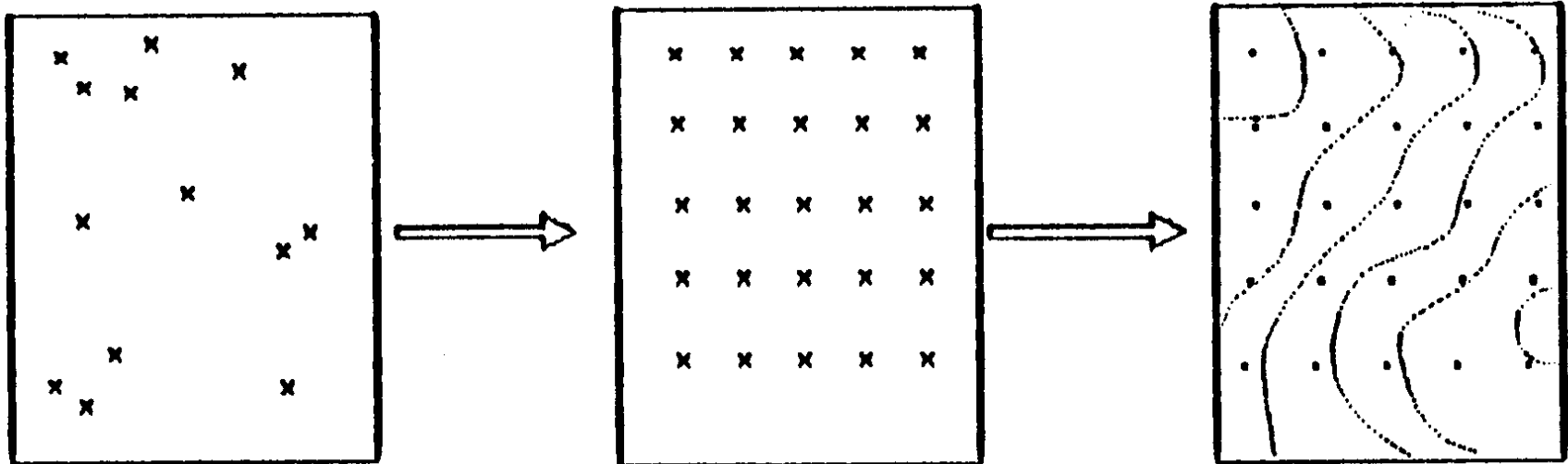
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Observational Data



A Popular Method for Ocean Data Analysis: Optimum Interpolation (OI)



OI – Equation

Grid point $\rightarrow k$, Observational Point $\rightarrow j$

$Q_k^f \rightarrow$ First guess field (gridded)

$Q_j^o \rightarrow$ Observation

$Q_j^f \rightarrow$ First guess interpolated on the observational point

$$Q_k^a = Q_k^f + \sum_{j=1}^N a_{kj} (Q_j^o - Q_j^f)$$

$Q_k^a \rightarrow$ Analyzed field at the grid point

OI – Weight Coefficients a_{kj}

$$\sum_{j=1}^N (h_{ij} + d_{ij} I_i^o) a_{kj} = h_{kj}$$

h_{ij} h_{kj} \rightarrow Autocorrelation functions

I_i^o \rightarrow Signal-to-noise ratio

Three Requirements for the OI Method

- (1) First guess field
- (2) Autocorrelation functions
- (3) High signal-to-noise ratio

What happens if the three
conditions are not
satisfied?

Spectral Representation - a Possible Alternative Method

$$c(\mathbf{x}, z_k, t) = A_0(z_k, t) + \sum_{m=1}^M A_m(z_k, t) \Psi_m(\mathbf{x}, z_k),$$

$\Psi_m \rightarrow$ Basis functions

$c \rightarrow$ any ocean variable

Flow Decomposition

$$u = \frac{\partial \Psi}{\partial y} + \frac{\partial^2 \Phi}{\partial x \partial z}, \quad v = -\frac{\partial \Psi}{\partial x} + \frac{\partial^2 \Phi}{\partial y \partial z},$$

$$\Delta \Psi = -\zeta$$

$$\Delta \Phi = -w$$

Basis Functions (Closed Basin)

$$\Delta \Psi_k = -\lambda_k \Psi_k, \quad \Psi_k|_{\Gamma} = 0, \quad k = 1, \dots, \infty$$

$$\Delta \Phi_m = -\mu_m \Phi_m, \quad \frac{\partial \Phi_m}{\partial n}|_{\Gamma} = 0, \quad m = 1, \dots, \infty.$$

Basis Functions (Open Boundaries)

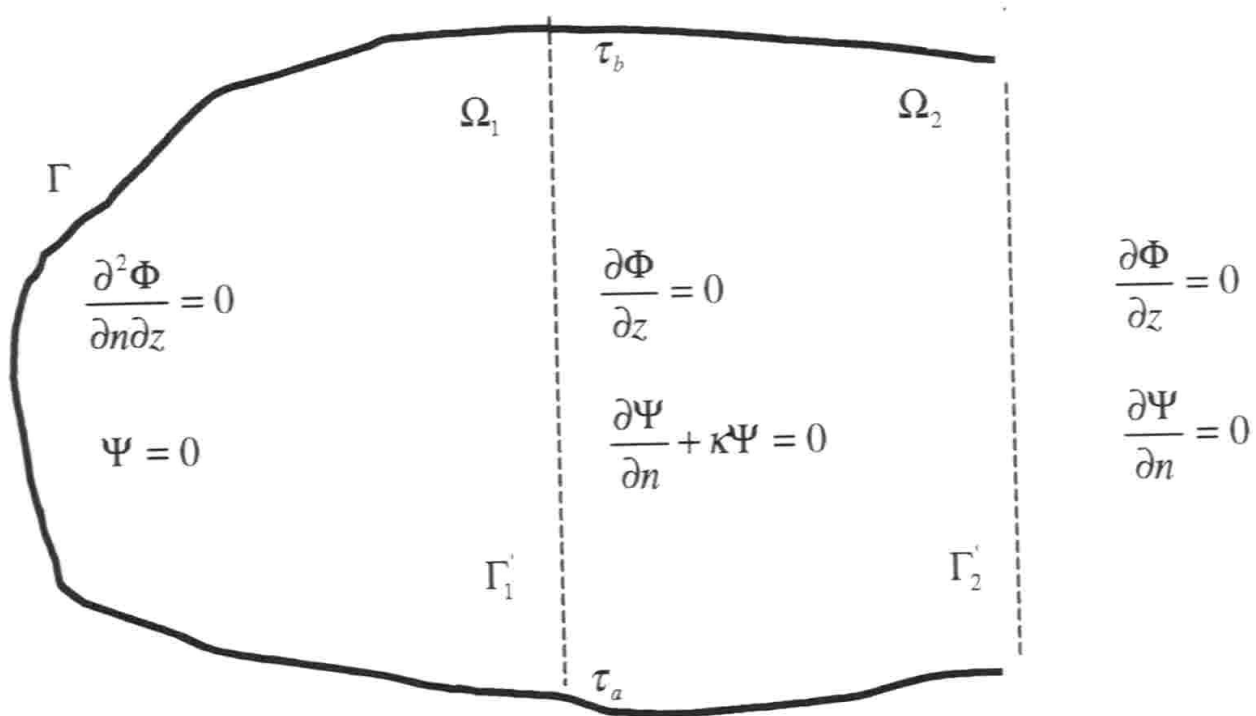
$$\Delta \Psi_k = -\lambda_k \Psi_k,$$

$$\Delta \Phi_m = -\mu_m \Phi_m,$$

$$\Psi_k|_{\Gamma} = 0, \quad \frac{\partial \Phi_m}{\partial n}|_{\Gamma} = 0,$$

$$\left[\frac{\partial \Psi_k}{\partial n} + \kappa(\tau) \Psi_k \right] |_{\Gamma'_1} = 0, \quad \Phi_m|_{\Gamma'_1} = 0,$$

Boundary Conditions



Spectral Decomposition

$$u_{KM} = \sum_{k=1}^K a_k(z, t^\circ) \frac{\partial \Psi_k(x, y, z, \kappa^\circ)}{\partial y} + \sum_{m=1}^M b_m(z, t^\circ) \frac{\partial \Phi_m(x, y, z)}{\partial x},$$

$$v_{KM} = - \sum_{k=1}^K a_k(z, t^\circ) \frac{\partial \Psi_k(x, y, z, \kappa^\circ)}{\partial x} + \sum_{m=1}^M b_m(z, t^\circ) \frac{\partial \Phi_m(x, y, z)}{\partial y}$$

$$T(\mathbf{x}, t) = T_0(\mathbf{x}) + \sum_{m=1}^M \text{?} c_m(t) F_m(\mathbf{x}, t)$$

$$S(\mathbf{x}, t) = S_0(\mathbf{x}) + \sum_{m=1}^M \text{?} d_m(t) F_m(\mathbf{x}, t)$$

Benefits of Using OSD

- (1) Don't need first guess field
- (2) Don't need autocorrelation functions
- (3) Don't require high signal-to-noise ratio
- (4) Basis functions are pre-determined before the data analysis.

Optimal Mode Truncation

$$J(a_1, \dots, a_K, b_1, \dots, b_M, \kappa, P) = \frac{1}{2} \left(\|u_p^{obs} - u_{KM}\|_P^2 + \|v_p^{obs} - v_{KM}\|_P^2 \right) \rightarrow \min,$$

Vapnik (1983) Cost Function

$$J_{emp} = J(a_1, \dots, a_K, b_1, \dots, b_M, \kappa, P).$$

$$\text{Prob} \left\{ \sup_{K, M, S} |\langle J(K, M, S) \rangle - J_{emp}(K, M, S)| \geq \mu \right\} \leq g(P, \mu)$$

$$\lim_{P \rightarrow \infty} g(P, \mu) = 0$$

Optimal Truncation

- Gulf of Mexico, Monterey Bay, Louisiana-Texas Shelf, North Atlantic
- $K_{\text{opt}} = 40$, $M_{\text{opt}} = 30$

Determination of Spectral Coefficients (Ill-Posed Algebraic Equation)

$$\mathbf{A} \hat{\mathbf{a}} = \mathbf{QY},$$

This is caused by the features of the matrix **A**.

Rotation Method (Chu et al., 2004)

$$\mathbf{SA} \hat{\mathbf{a}} = \mathbf{SQY},$$

$$J_1 = \|\mathbf{A}\|^2 - \frac{\|\mathbf{SQY}\|^2}{\|\mathbf{a}\|^2} \rightarrow \max,$$

Example-1

Temporal and spatial
variability of Pacific Ocean

T (10 m) 1990-2008



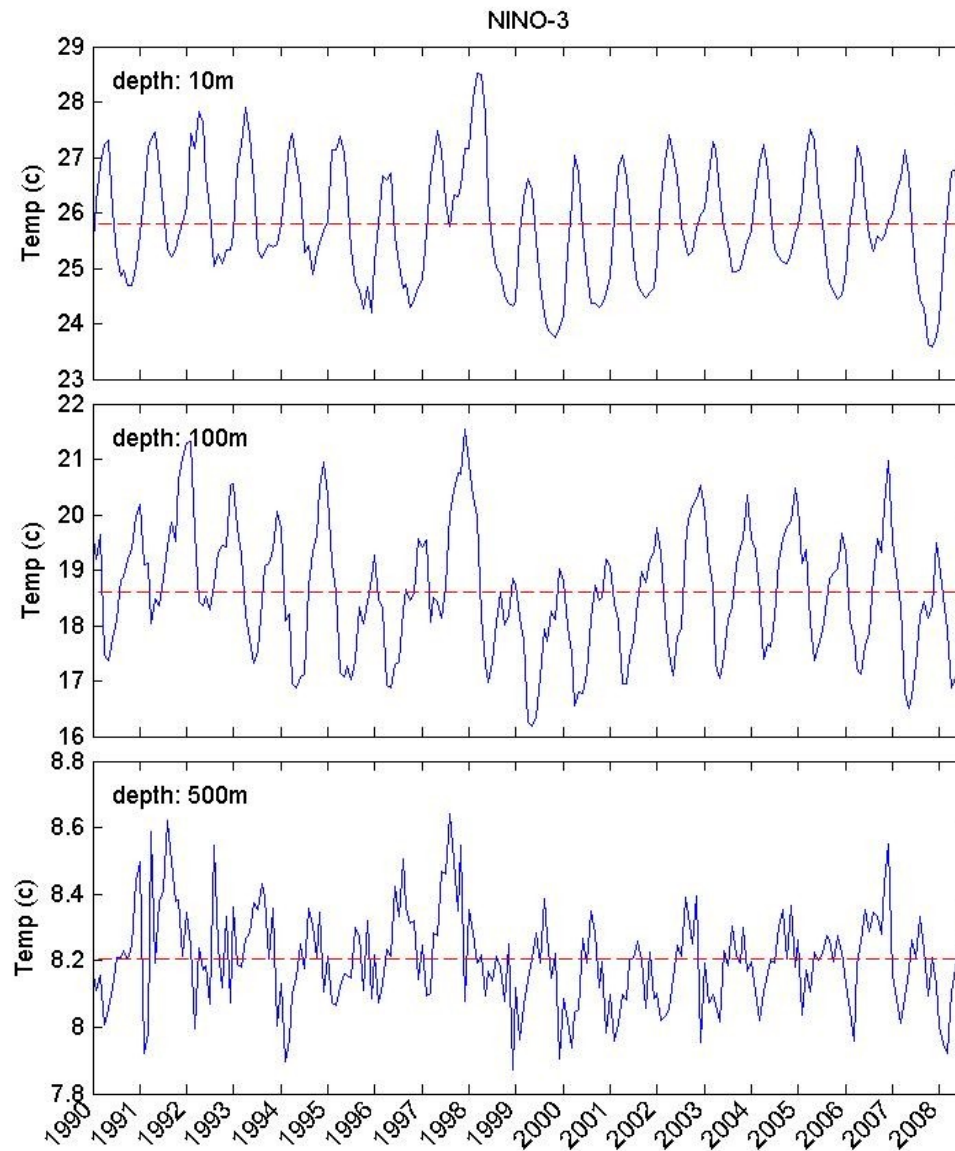
T (100 m) 1990-2008



T (500 m) 1990-2008



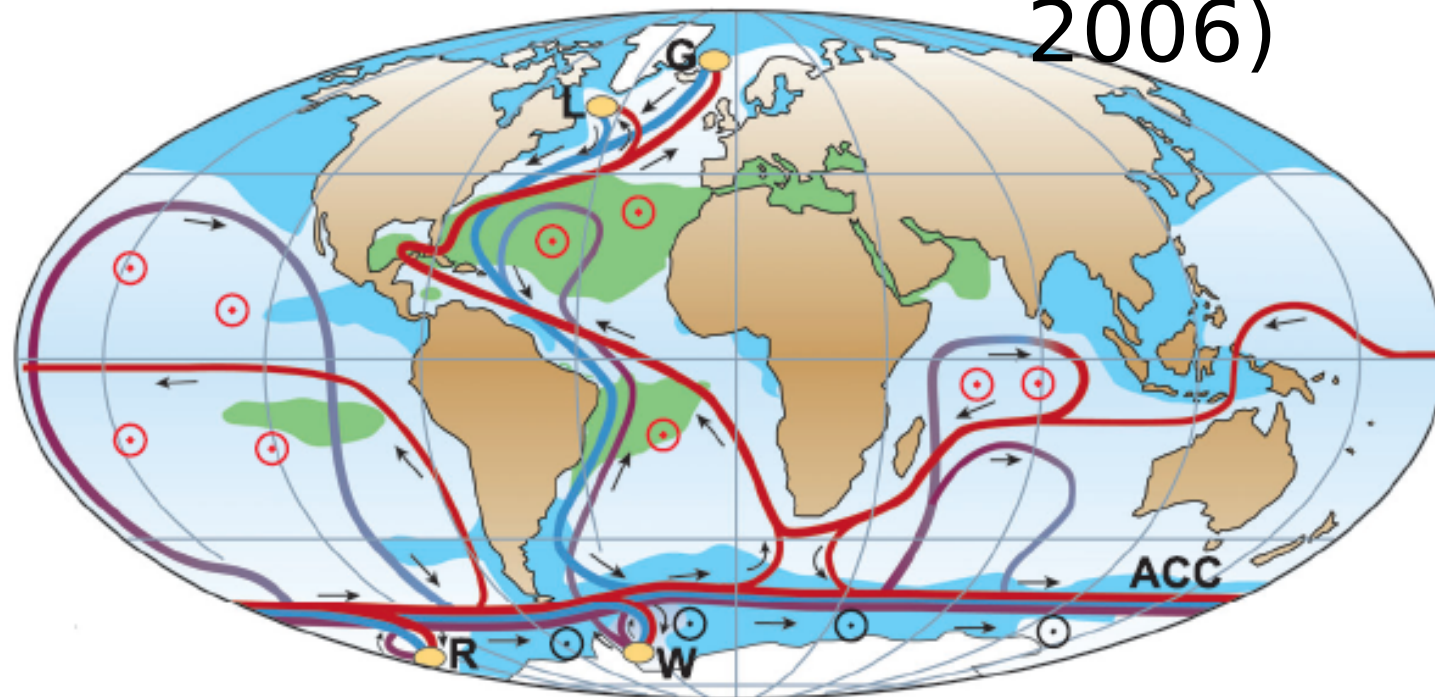
T: NINO-3 (5°S - 5°N , 150°W - 90°W)



Example-2 OSD for Analyzing ARGO Data

Baroclinic Rossby Waves in the tropical North
Atlantic

Tropical North Atlantic (4° - 24° N)
Important Transition Zone →
Meridional Overturning Circulation
(MOC) (Rahmstorf
2006)



- Surface flow
- Deep flow
- Bottom flow
- Deep Water Formation

- ⊙ Wind-driven upwelling
- ⊙ Mixing-driven upwelling
- Salinity > 36 ‰
- Salinity < 34 ‰

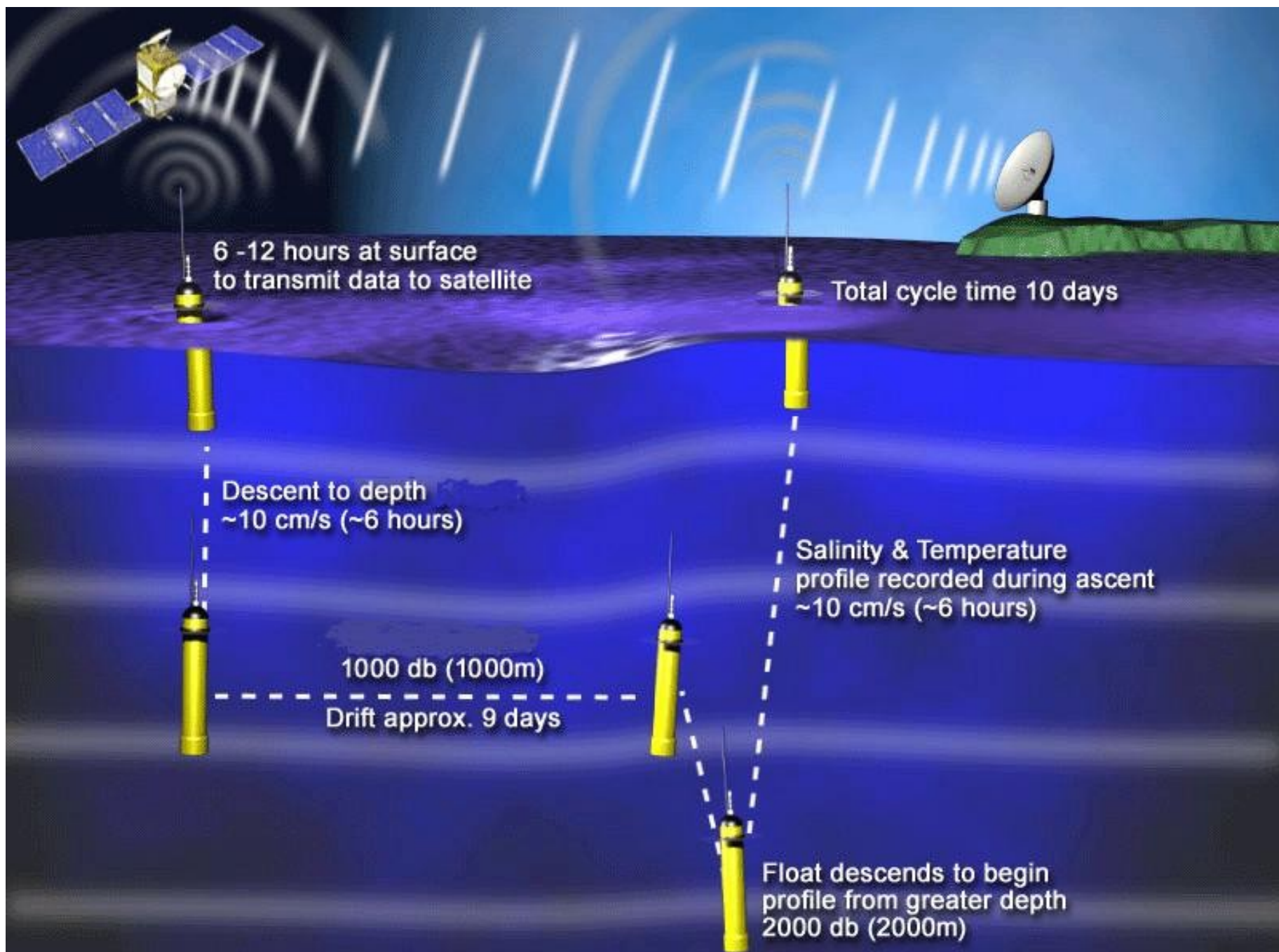
- L Labrador Sea
- G Greenland Sea
- W Weddell Sea
- R Ross Sea

MOC Variation →

Heat Transport Variation →

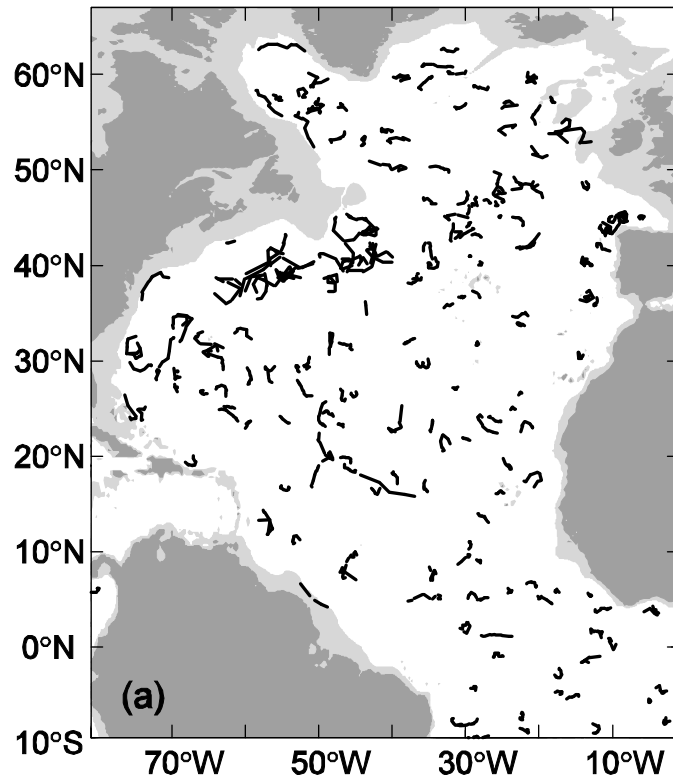
Climate Change

- Are mid-depth (~ 1000 m) ocean circulations steady?
- If not, what mechanisms cause the change? (Rossby wave propagation)

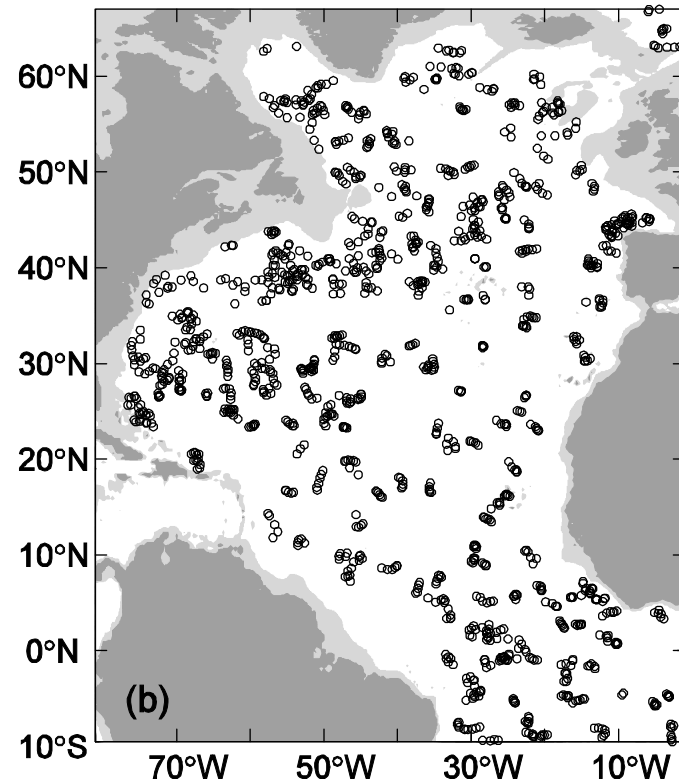


ARGO Observations (Oct-Nov 2004)

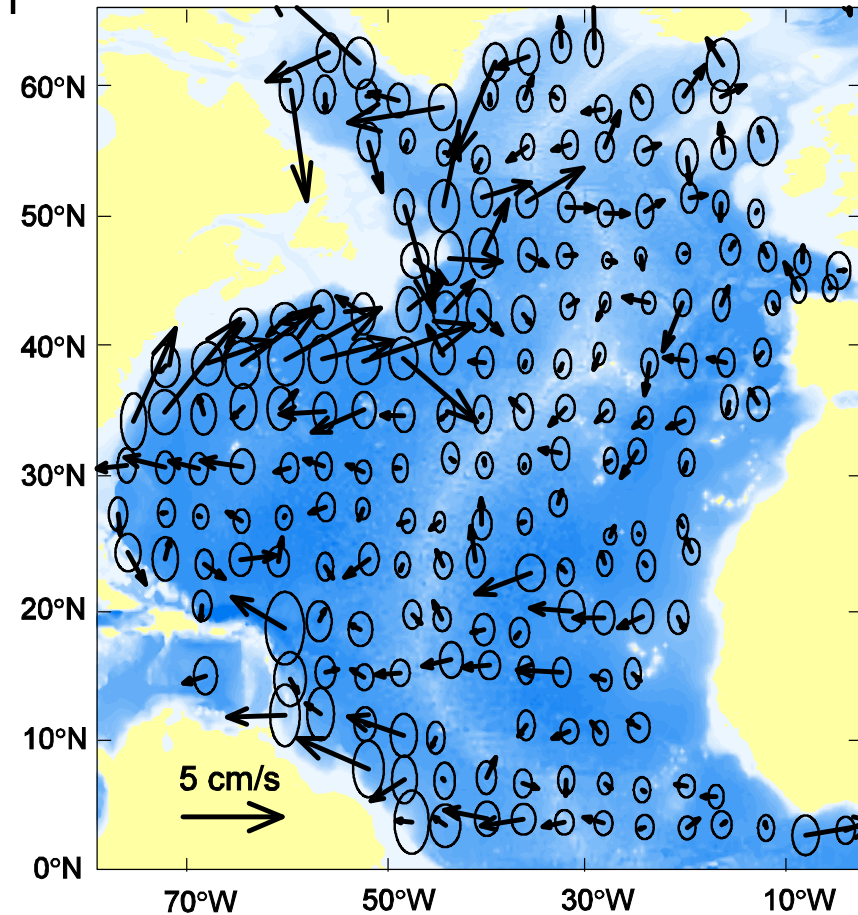
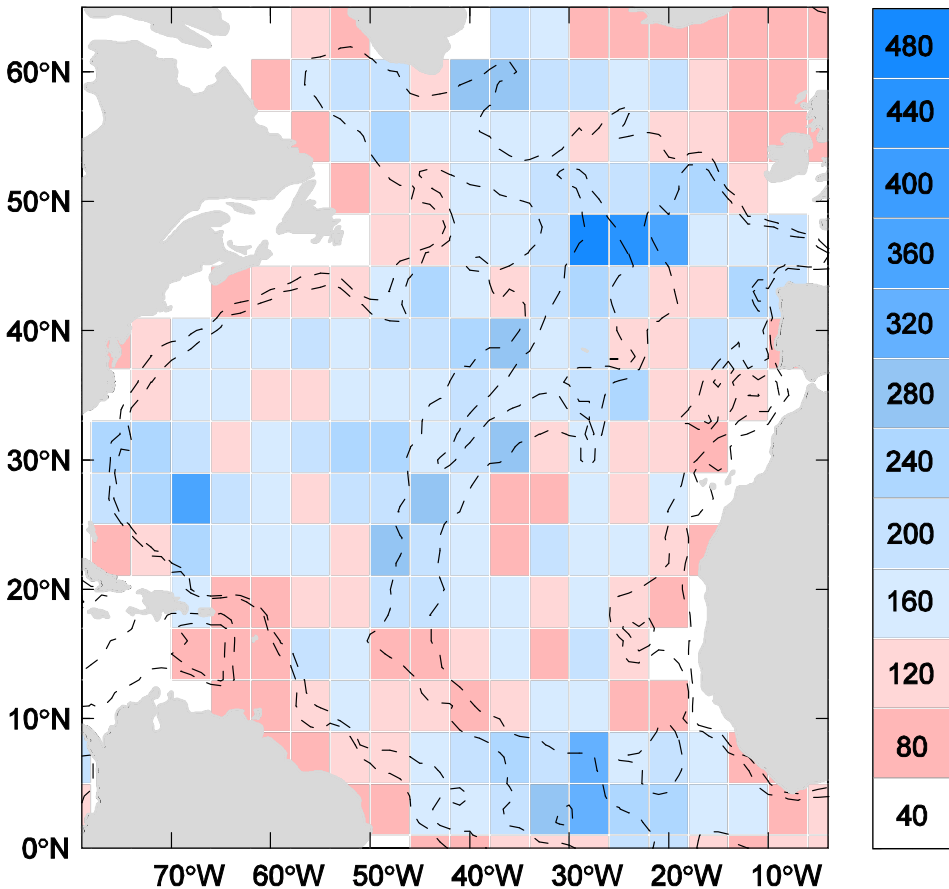
(a) Subsurface tracks



(b) Float positions where (T,S) were measured

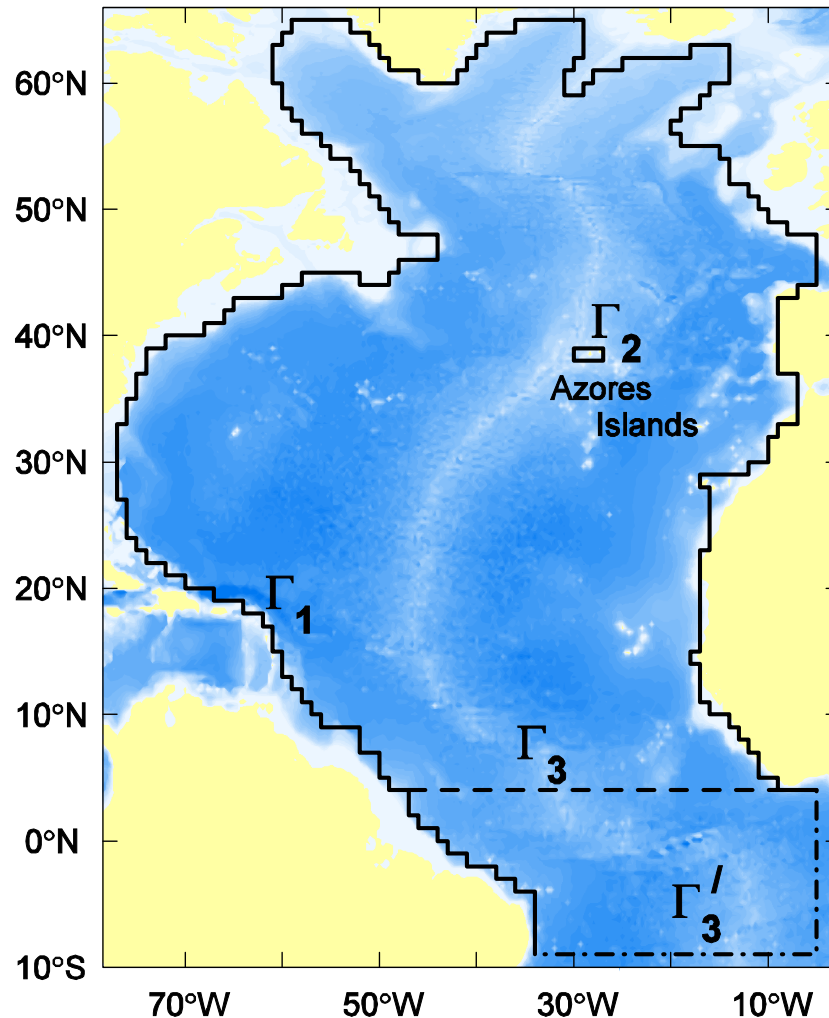


Circulations at 1000 m estimated from the original ARGO float tracks (bin method) April 2004 – April 2005

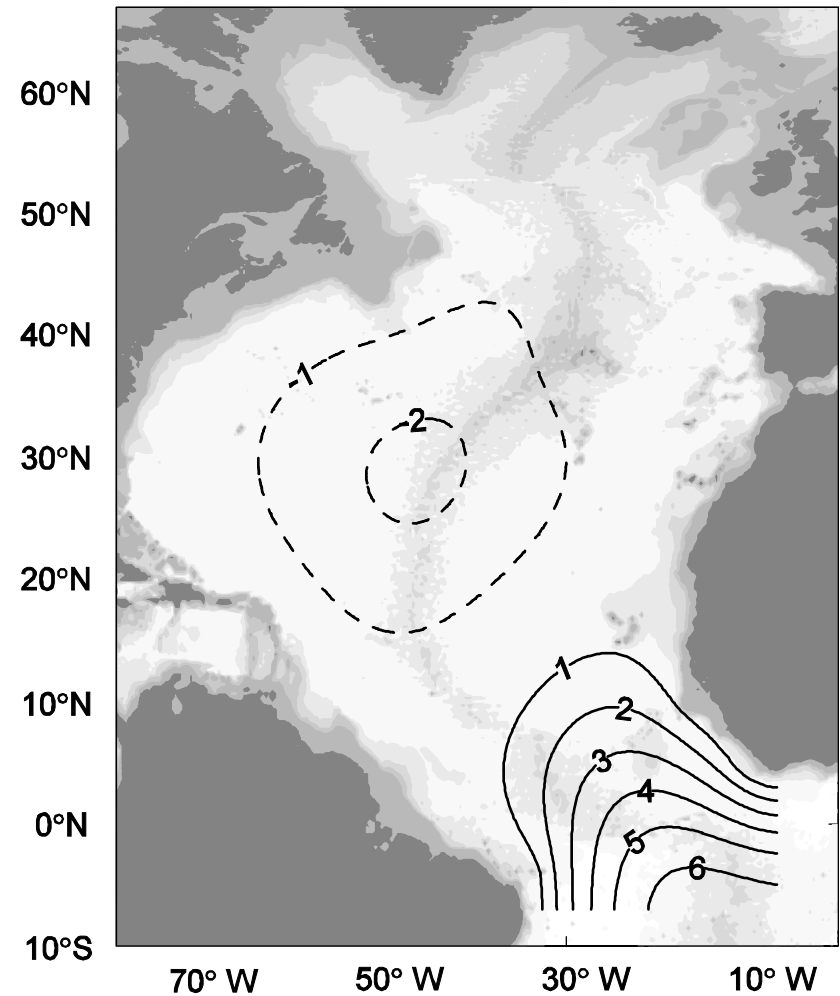
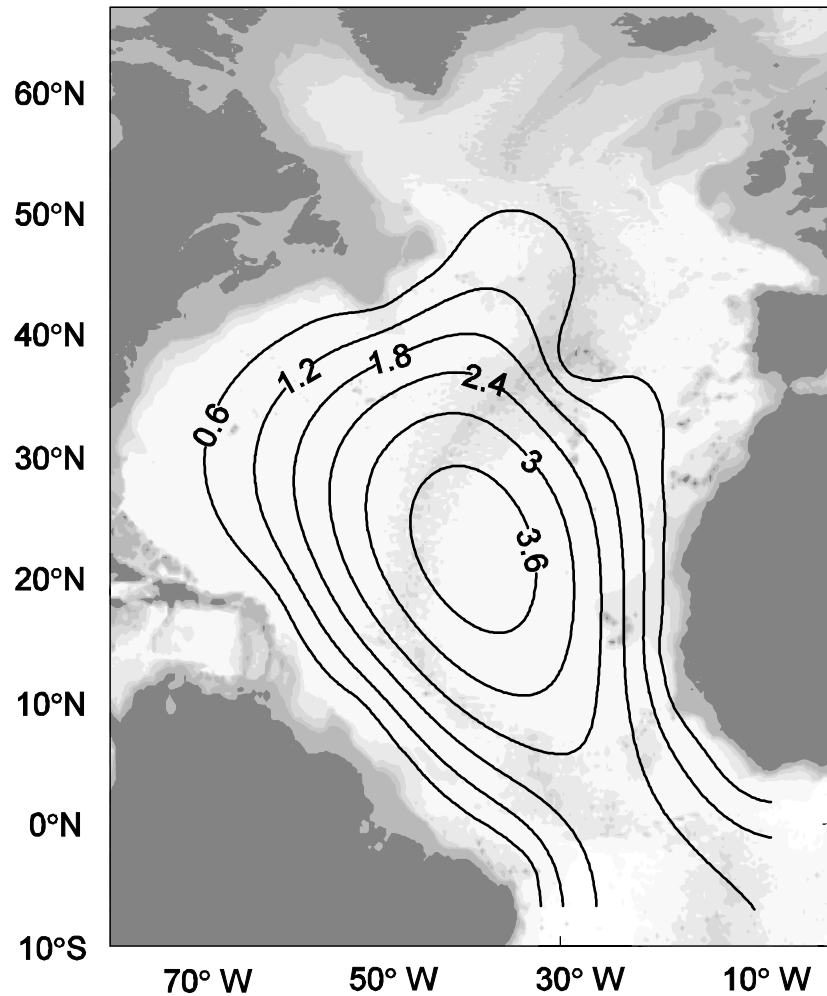


is **difficult** to get physical insights and
use such noisy data into ocean numerical models.

Boundary Configuration → Basis Functions for OSD

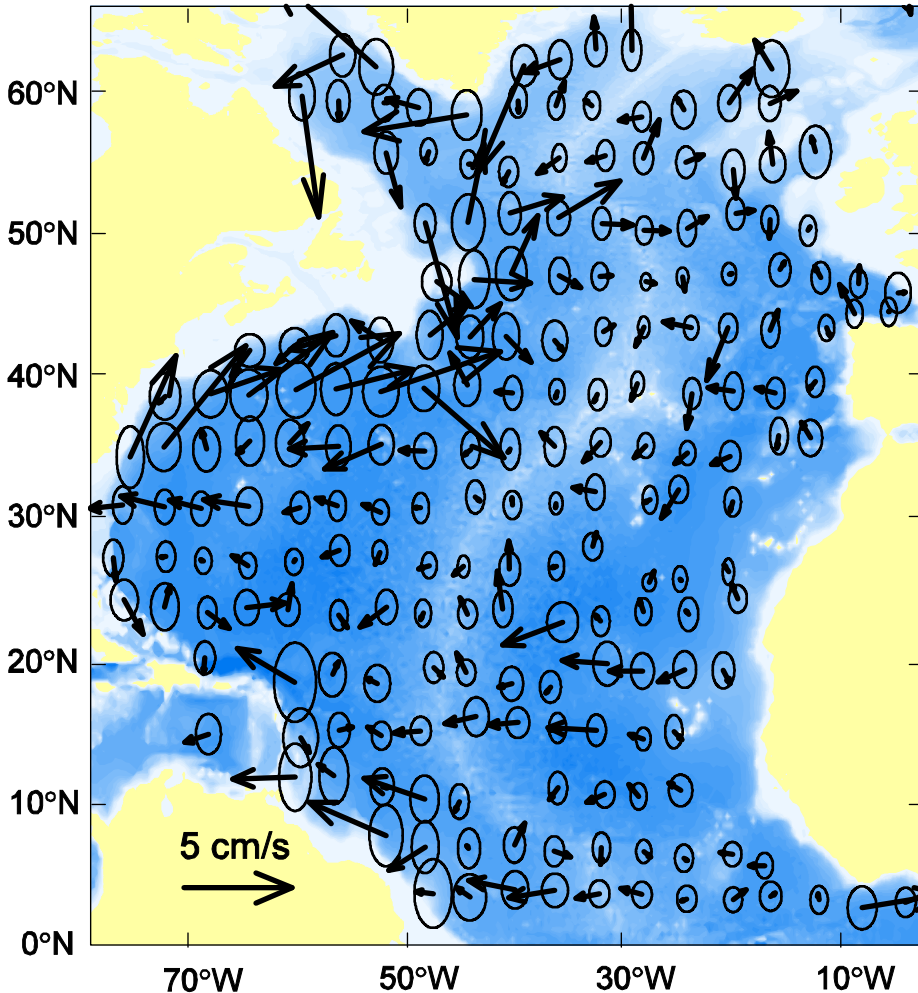


Basis Functions for Streamfunction Mode-1 and Mode-2

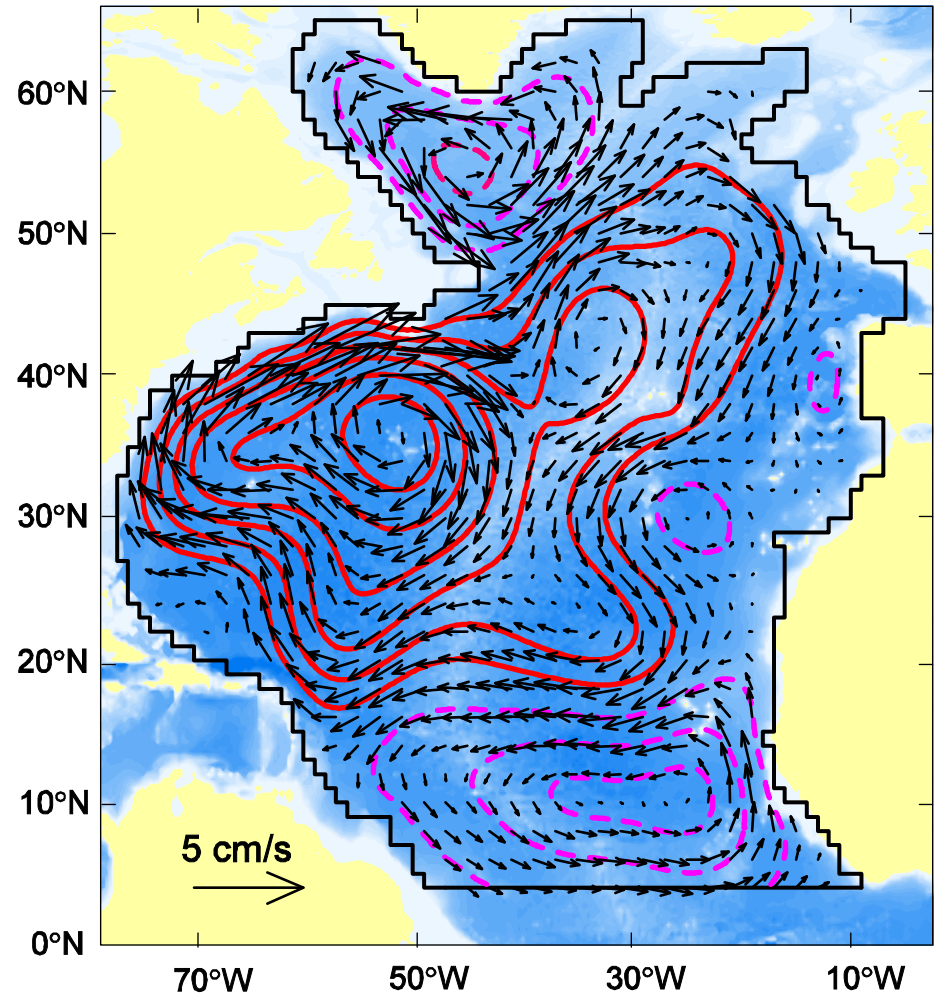


Circulations at 1000 m (March 04 to May 05)

Bin Method

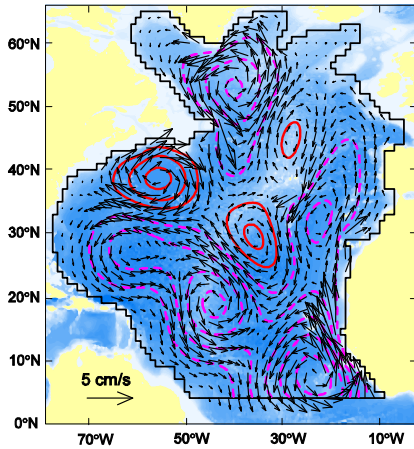


OSD

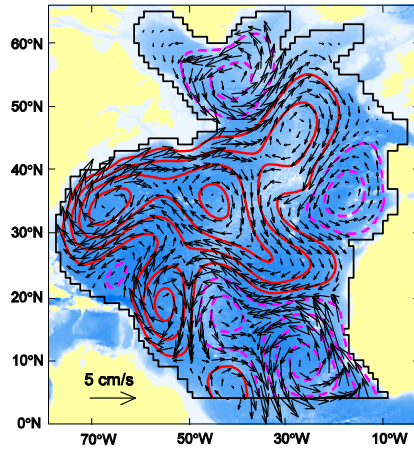


Mid-Depth Circulations (1000 m)

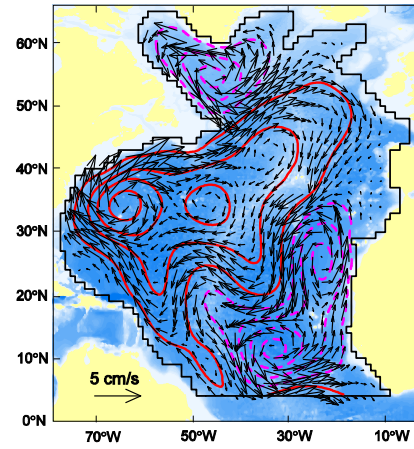
Mar-May 04



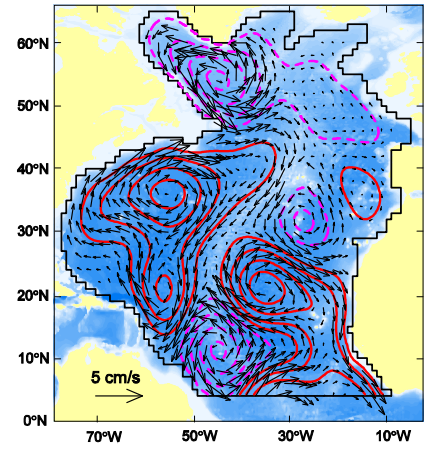
May - Jul 04



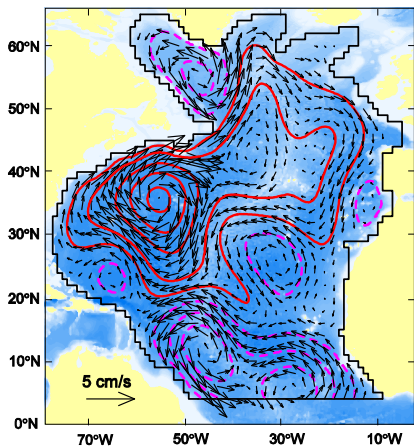
Jul-Sep 04



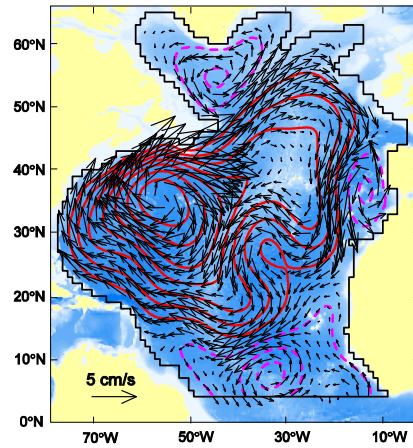
Sep - Nov 04



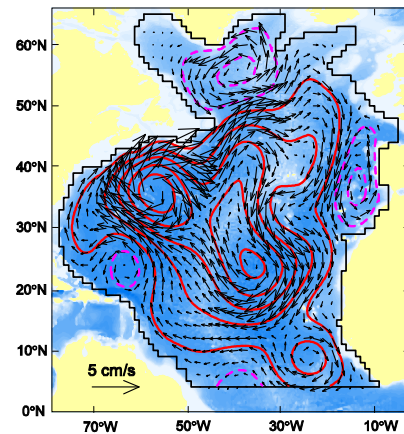
Nov 04 - Jan 05



Jan-Mar 05

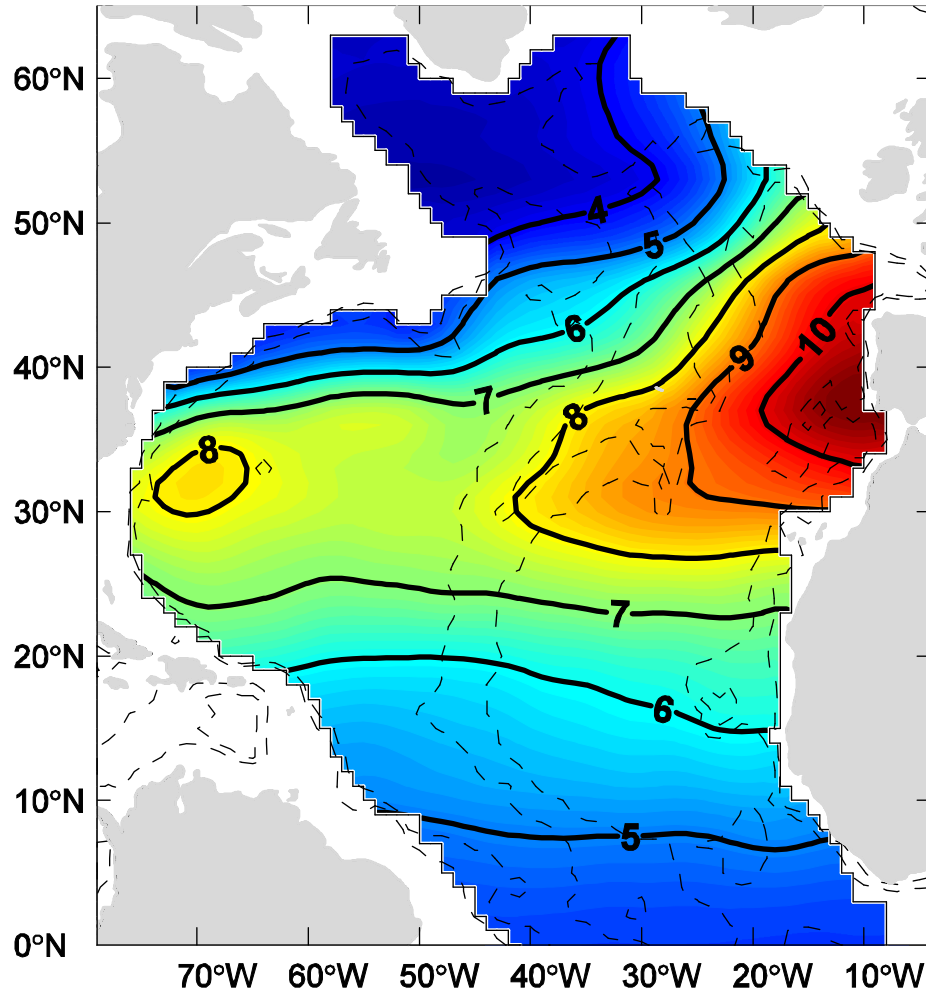


Mar - May 05

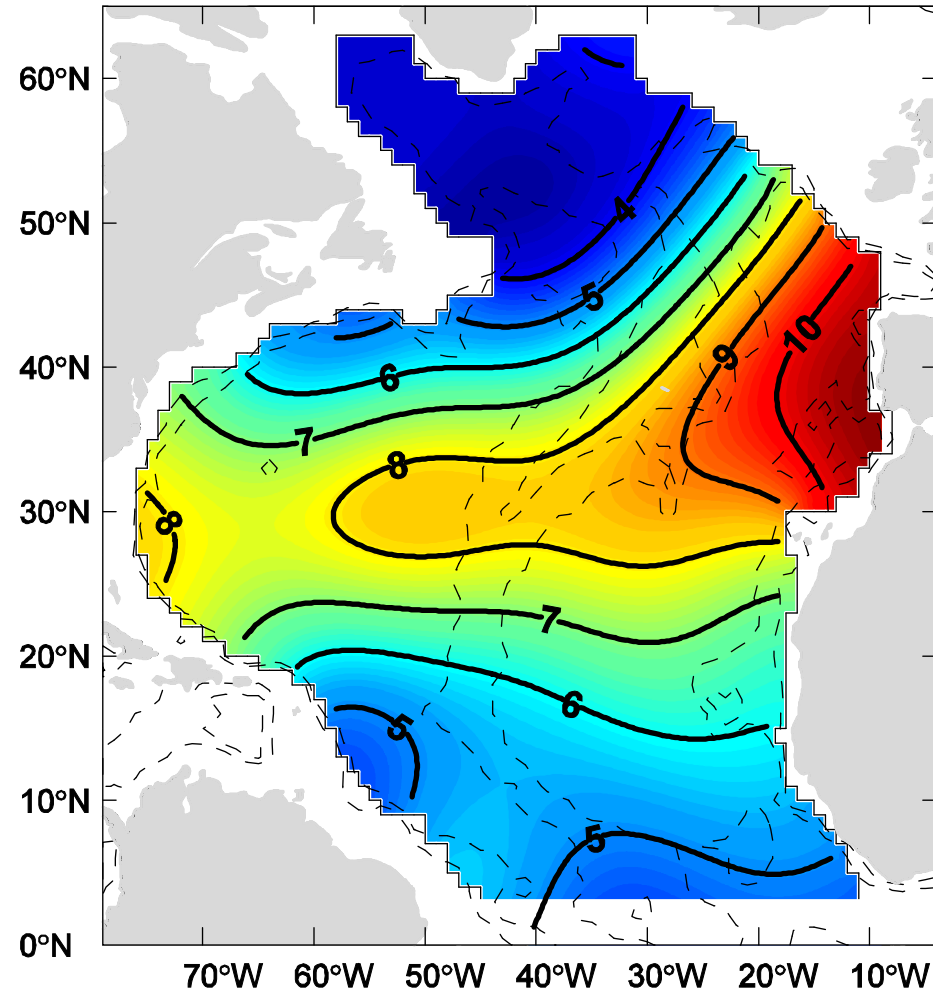


Temperature at 950 m (March 04 to May 05)

NOAA/WOA

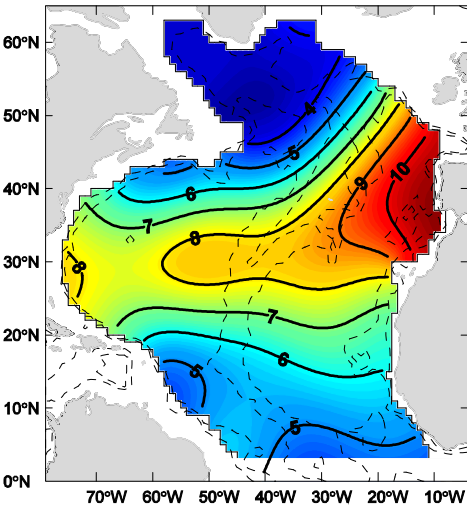


OSD

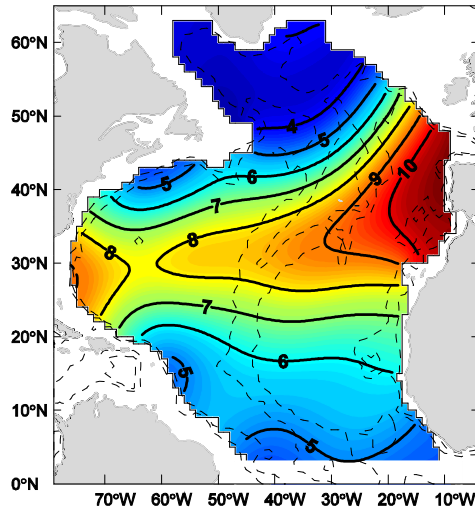


Mid-Depth Temperature (950 m)

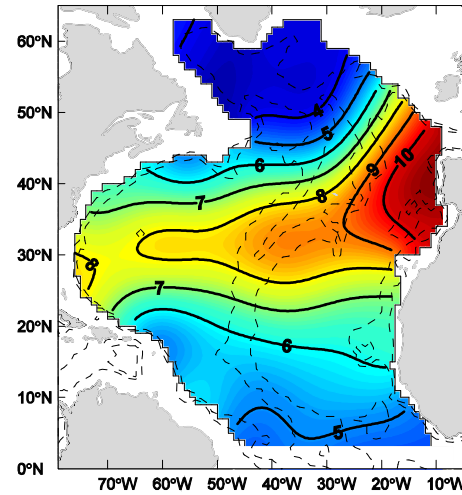
May 04



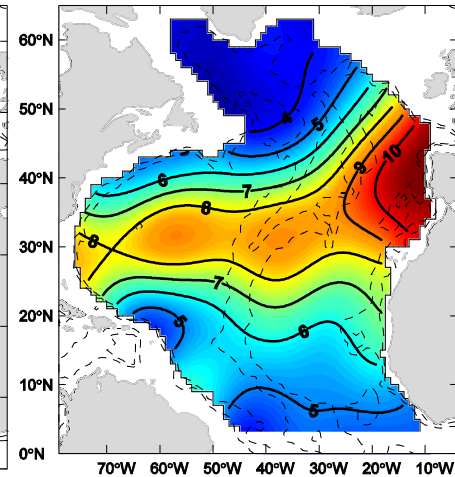
Jul 04



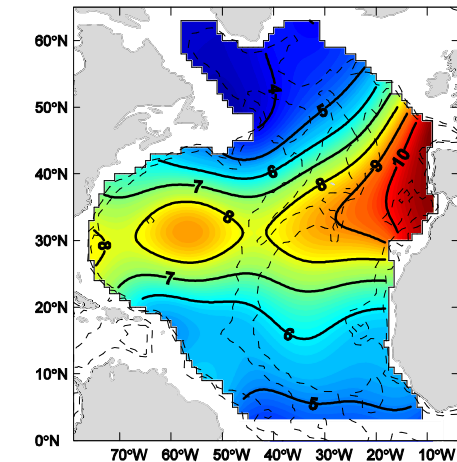
Sep 04



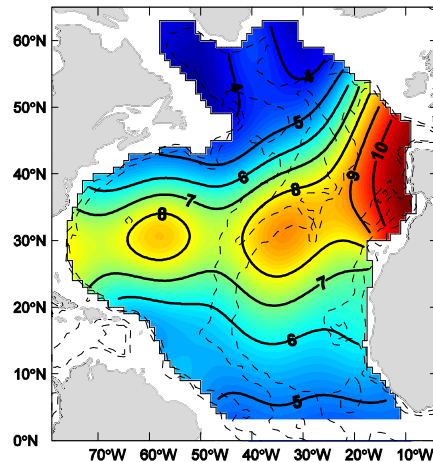
Nov 04



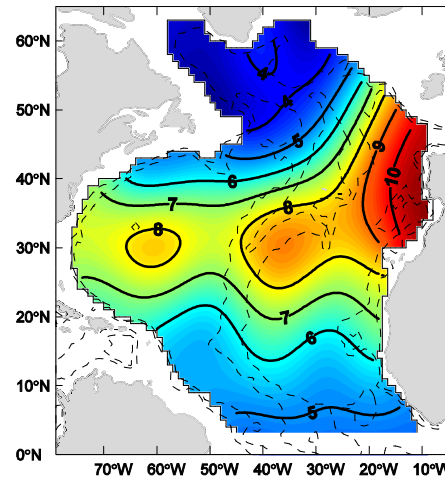
Jan 05



Mar 05



May 05



Baroclinic Rossby Waves in Tropical North Atlantic

Fourier Expansion → Temporal Annual and Semi-annual

$$\hat{\psi} \approx \bar{\psi}(\mathbf{x}_{\perp}) + \psi_1(\mathbf{x}_{\perp}, t) + \psi_2(\mathbf{x}_{\perp}, t),$$

$$\psi_1(\mathbf{x}_{\perp}, t) = \sum_{s=1}^2 A_{\omega_1, s} \cos(\omega_1 t + \theta_{\omega_1, s}) Z_s(\mathbf{x}_{\perp}) + \sum_{k=1}^{K_{opt}} B_{\omega_1, k} \cos(\omega_1 t + \vartheta_{\omega_1, k}) \Psi_k(\mathbf{x}_{\perp}),$$

$$\psi_2(\mathbf{x}_{\perp}, t) = \sum_{s=1}^2 A_{\omega_2, s} \cos(\omega_2 t + \theta_{\omega_2, s}) Z_s(\mathbf{x}_{\perp}) + \sum_{k=1}^{K_{opt}} B_{\omega_2, k} \cos(\omega_2 t + \vartheta_{\omega_2, k}) \Psi_k(\mathbf{x}_{\perp}),$$

$$T_0 = 12 \text{ months}; \quad \omega_1 = 2\pi / T_0 ; \quad \omega_2 = 4\pi / T_0$$

Fourier Expansion → Temporal Annual and Semi-annual

$$\hat{T}(\mathbf{x}_{\perp}, z, t) \approx \bar{T}(\mathbf{x}_{\perp}, z) + T_1(\mathbf{x}_{\perp}, z, t) + T_2(\mathbf{x}_{\perp}, z, t),$$

$$T_1(\mathbf{x}_{\perp}, z, t) = \sum_{m=1}^{M_{opt}} C_{\omega_1, m}(z) \cos[\omega_1 t + \chi_{\omega_1, m}(z)] \Xi_m(\mathbf{x}_{\perp}, z),$$

$$T_2(\mathbf{x}_{\perp}, z, t) = \sum_{m=1}^{M_{opt}} C_{\omega_2, m}(z) \cos[\omega_2 t + \chi_{\omega_2, m}(z)] \Xi_m(\mathbf{x}_{\perp}, z),$$

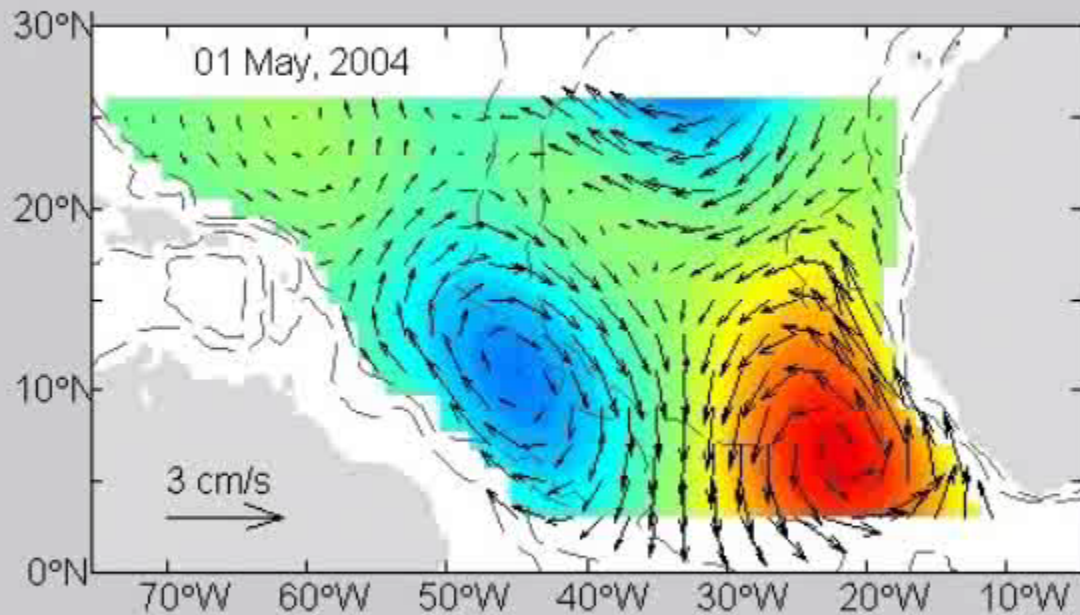
$$T_0 = 12 \text{ months}; \quad \omega_1 = 2\pi / T_0 ; \quad \omega_2 = 4\pi / T_0$$

Optimization

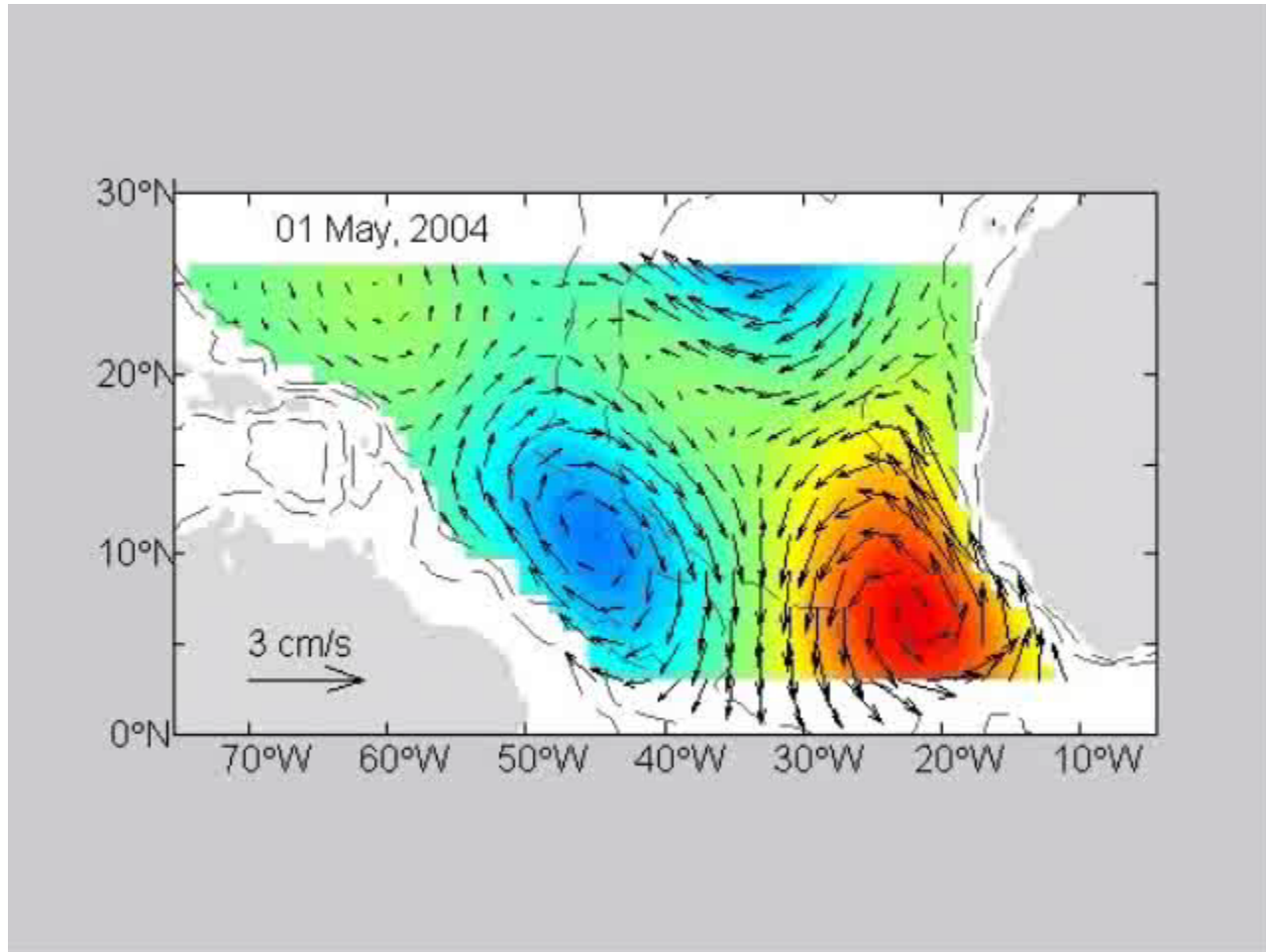
$$J_s = \int_{t_o}^{t_o + T_o} \left[a_s(t) - \sum_{\omega=\omega_1, \omega_2} A_{\omega,s} \cos(\omega t + \theta_{\omega,s}) \right]^2 dt \rightarrow \min$$

$$I_k = \int_{t_o}^{t_o + T_o} \left[b_k(t) - \sum_{\omega=\omega_1, \omega_2} B_{\omega,s} \cos(\omega t + \vartheta_{\omega,s}) \right]^2 dt \rightarrow \min$$

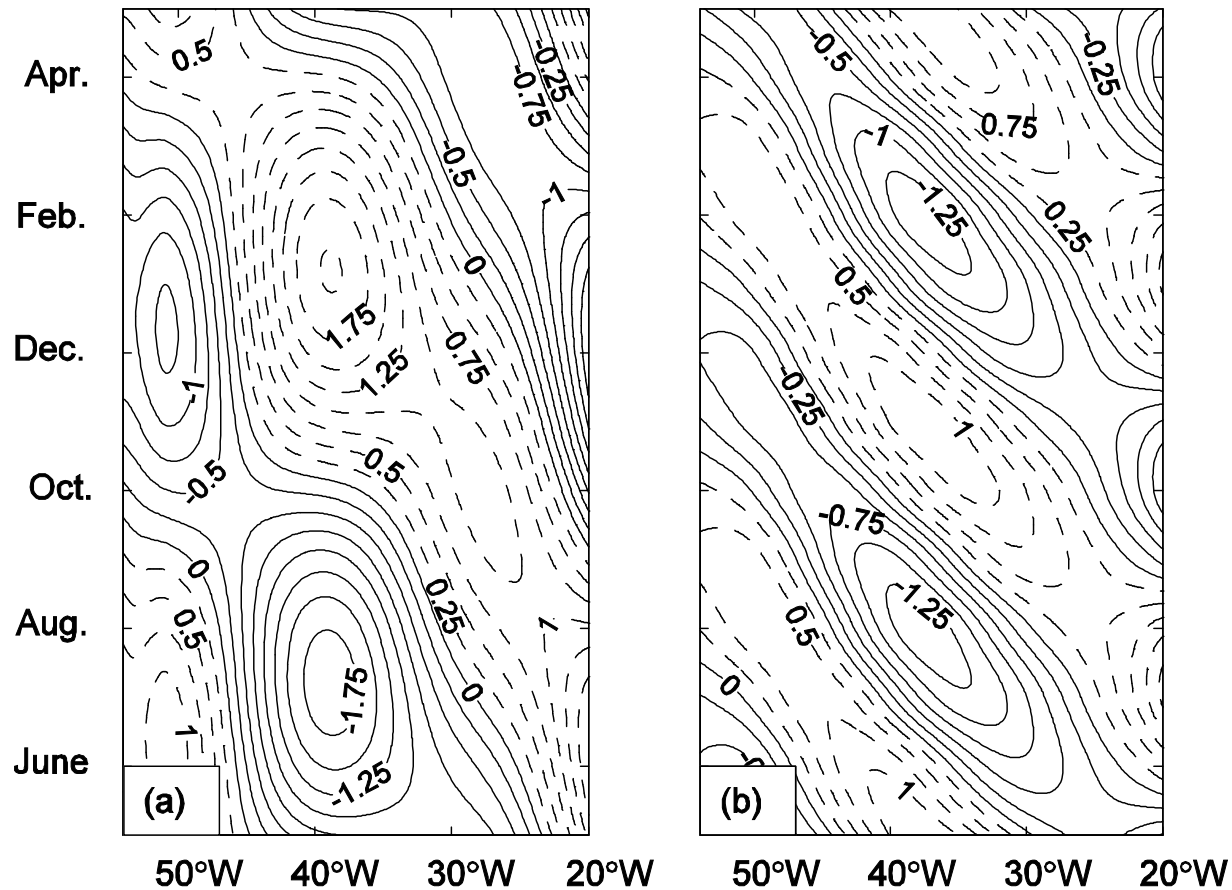
Annual Component



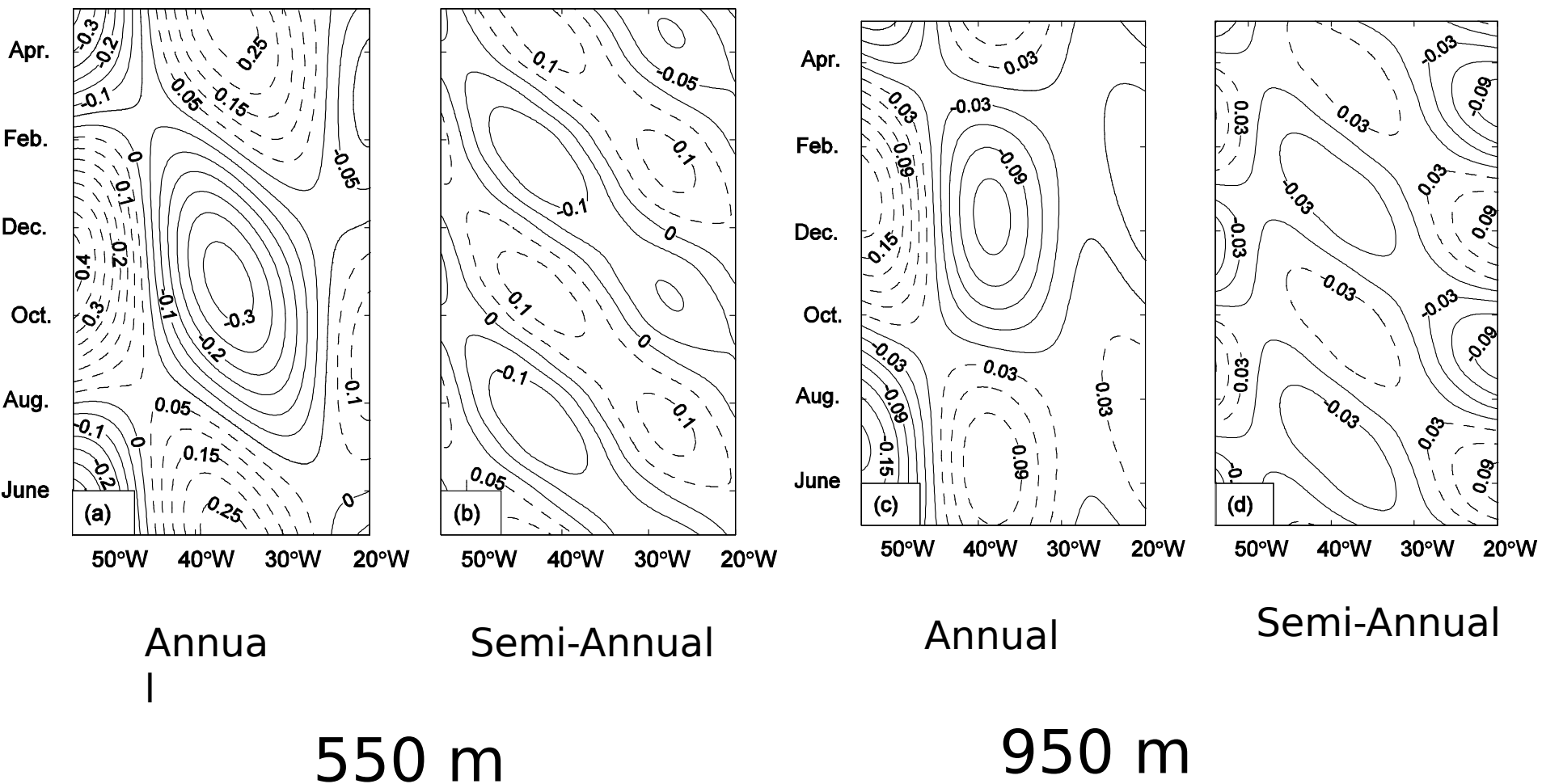
Semi-annual Component



Time -Longitude Diagrams of Meridional Velocity Along 11°N

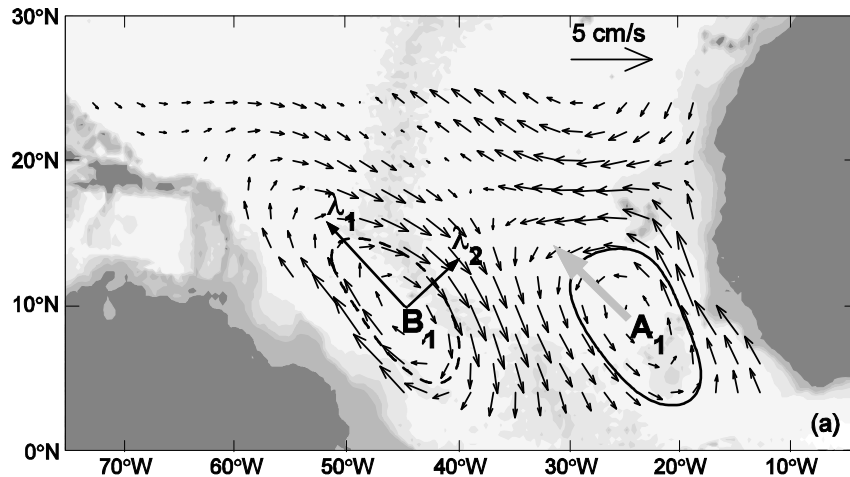


Time -Longitude Diagrams of temperature Along 11°N

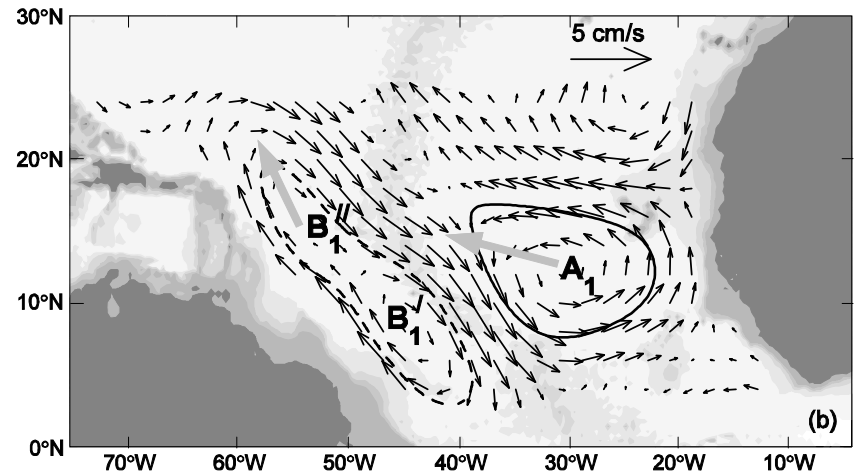


Annual Currents (1000 m)

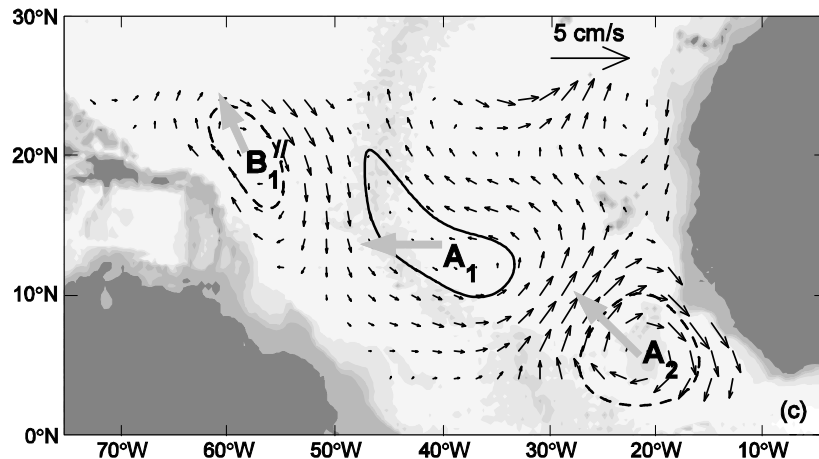
May-Jun 2004



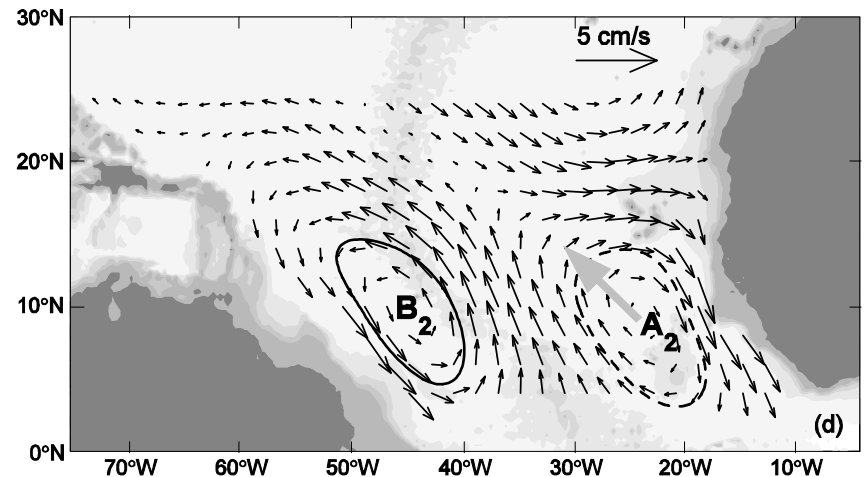
Jul-Aug 2004



Sep-Oct 2004



Nov-Dec 2004



Characteristics of Annual Rossby Waves

	March, 04 – May, 05 float data			March, 04 – May, 06 float data		
	c_p (cm/s)	L_1 (km)	L_2 (km)	c_p (cm/s)	L_1 (km)	L_2 (km)
5°N	12	1200	1100	12	1300	900
8°N	16	2500	1400	12	2100	1100
11°N	14	2200	1400	11	1900	1100
13°N	11	2100	1500	10	2300	1500

Western
Basin

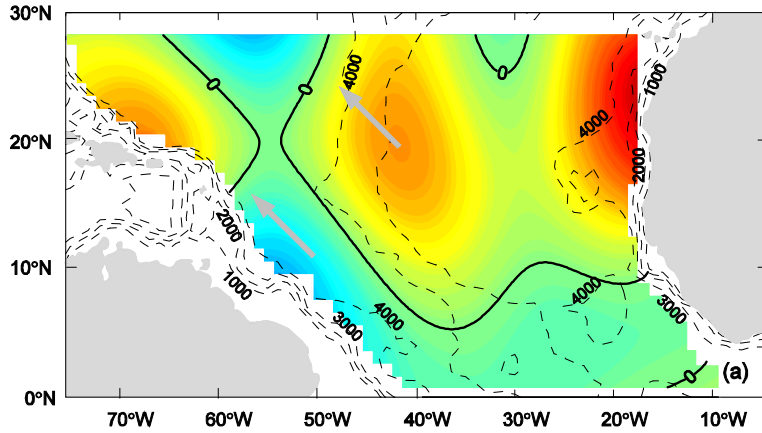
Eastern
Basin

Western
Basin

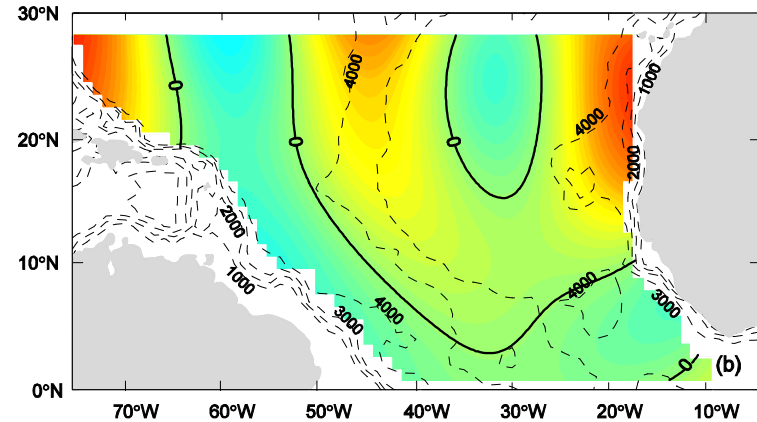
Eastern
Basin

Annual Monthly Temperature Anomaly ($^{\circ}\text{C}$) at 950 m Depth - Annual Rossby Waves (7-10 cm/s)

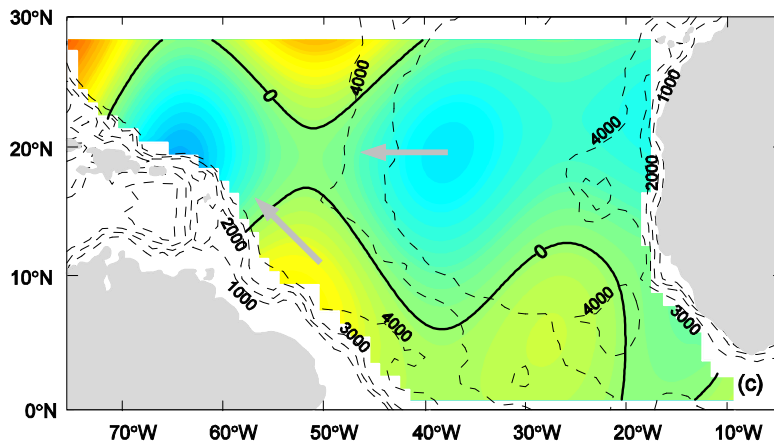
Jun 04



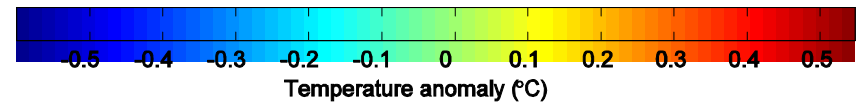
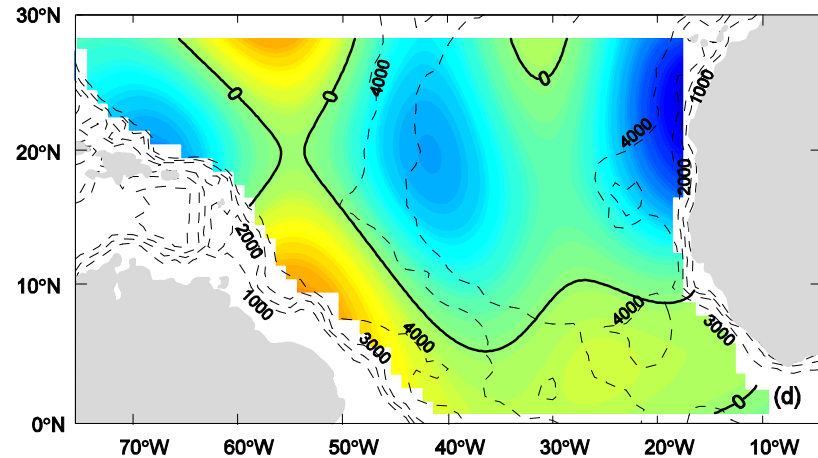
Aug 04



Oct 04

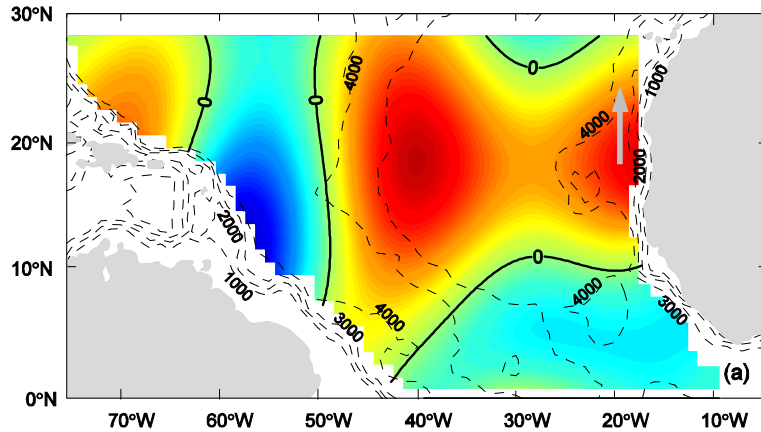


Dec 04

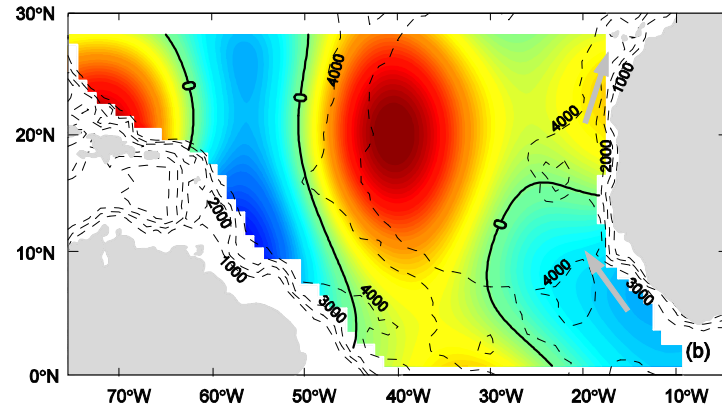


Annual Monthly Temperature Anomaly ($^{\circ}\text{C}$) at 250 m Depth - Equatorially Forced Coastal Kelvin waves (27-30 cm/s)

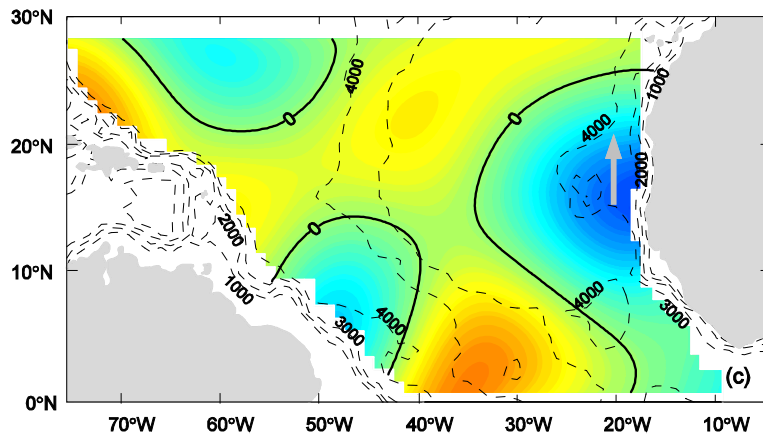
Jun 04



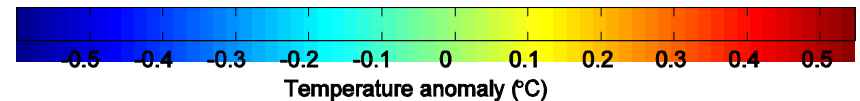
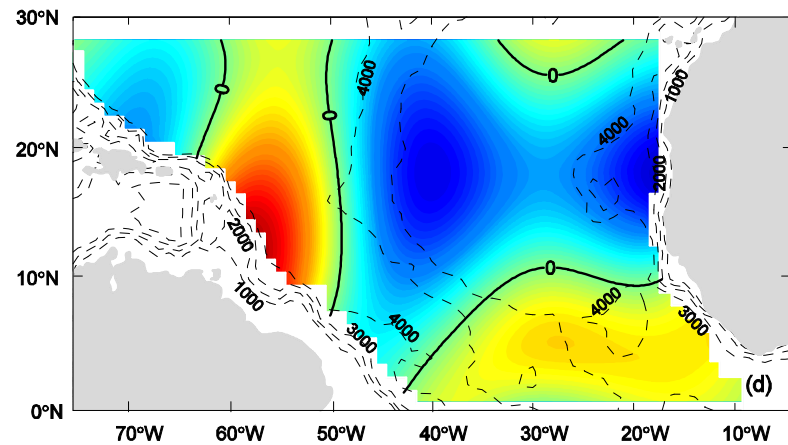
Aug 04



Oct 04

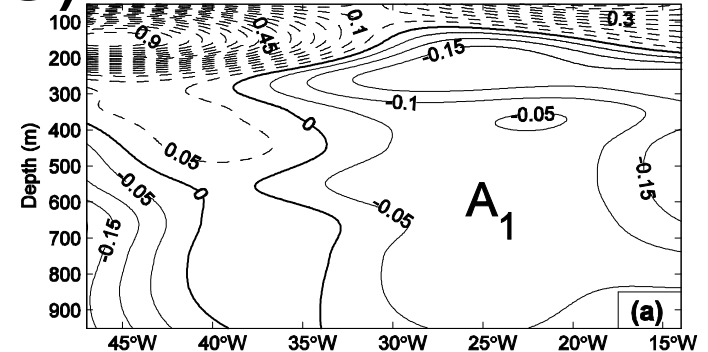


Dec 04

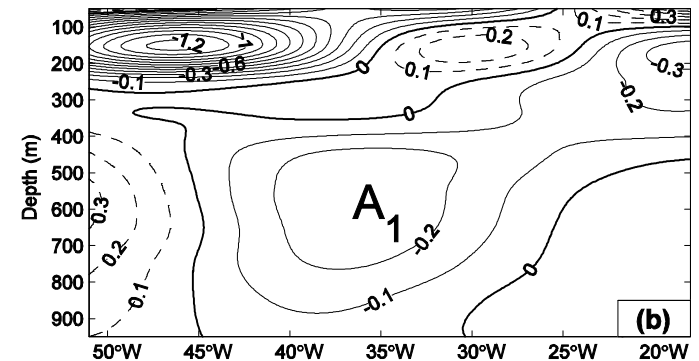


Zonal cross-sections of the annual component of the temperature anomaly ($^{\circ}\text{C}$)

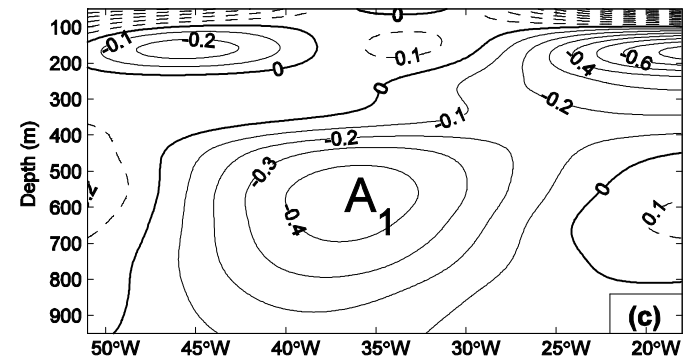
6°N in Jun 04 →



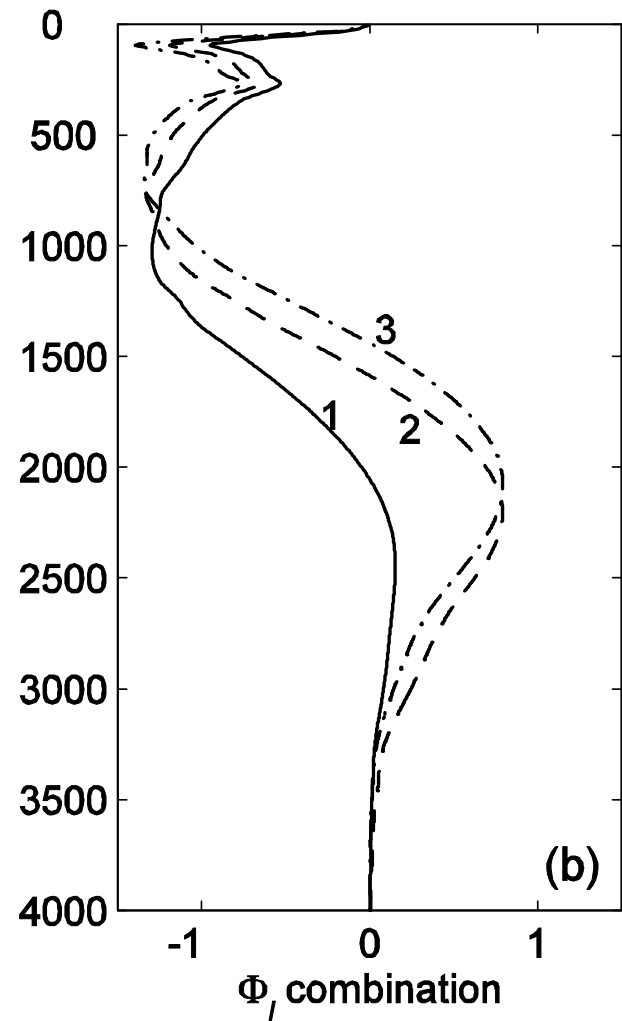
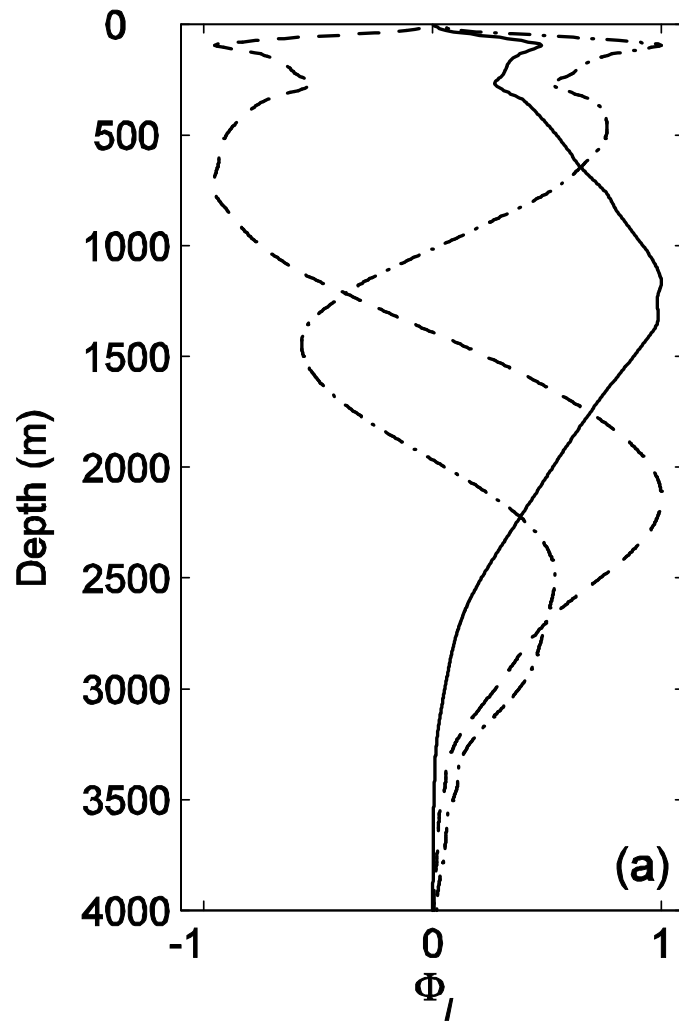
11°N in Oct, 04 →



16°N in Oct 04 →

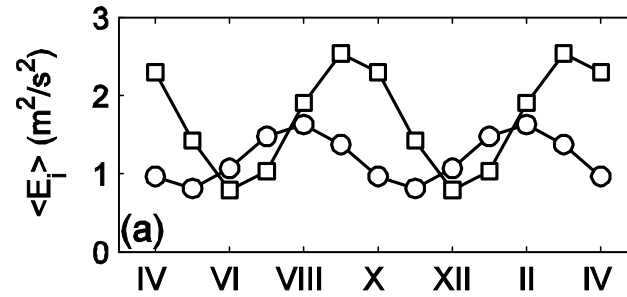


Baroclinic Modes



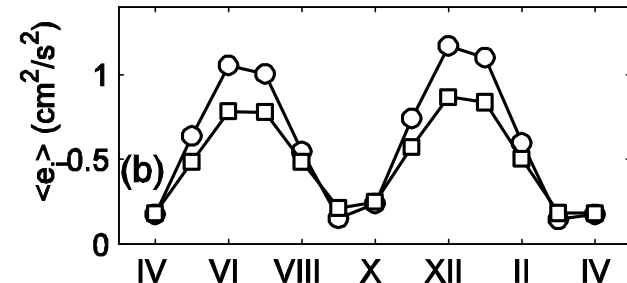
Annual Component in the Western Sub-Basin

Mean wind KE



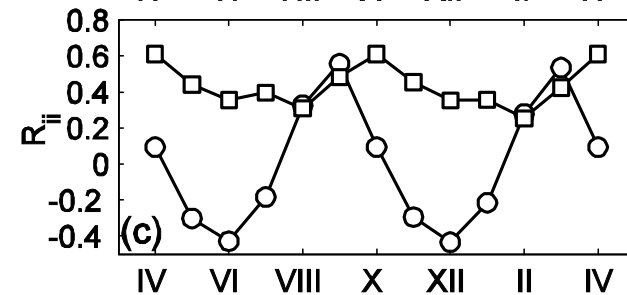
Zonal: circle

Mean KE for mid-depth currents

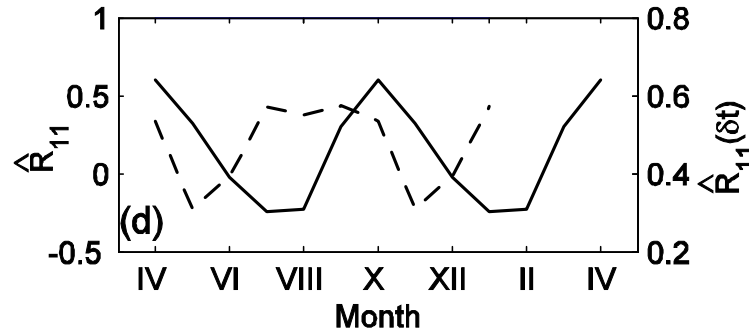


Meridional :
square

Correlation between Winds and currents

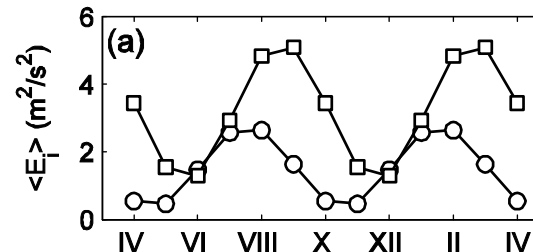


Correlation between wind Stress curl and streamfunction (solid: no-lag, dashed: 3 mon lag)



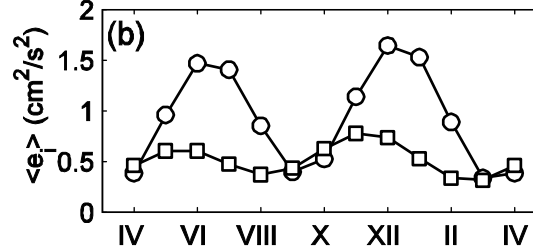
Annual Component in the Eastern Sub-Basin

Mean wind KE



Zonal: circle

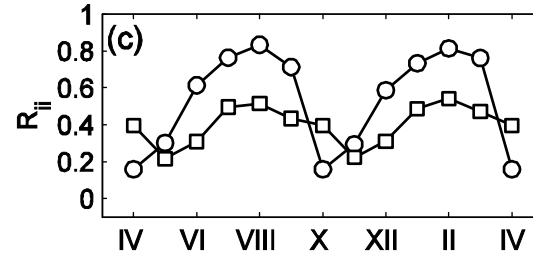
Mean KE for mid-depth currents



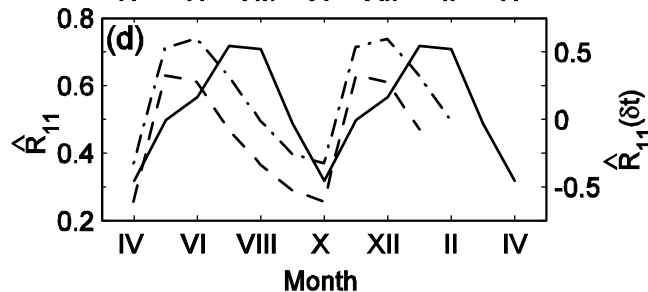
Meridional :

square

Correlation between Winds and currents

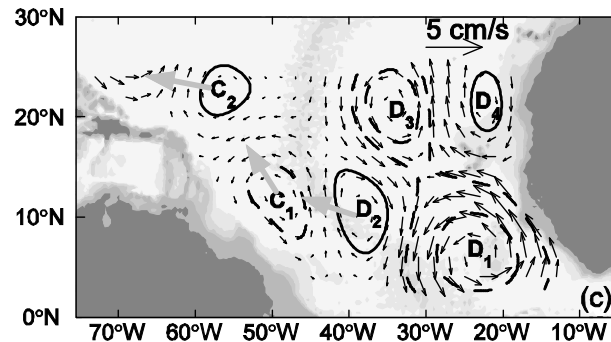
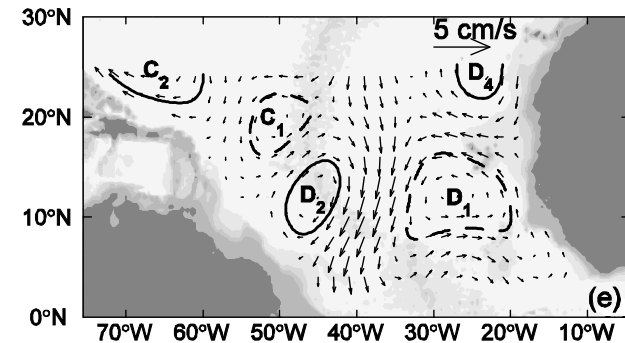
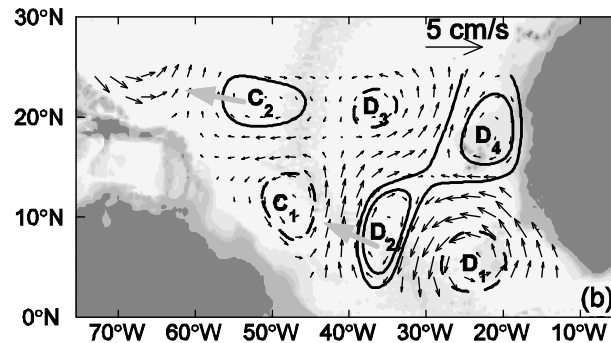
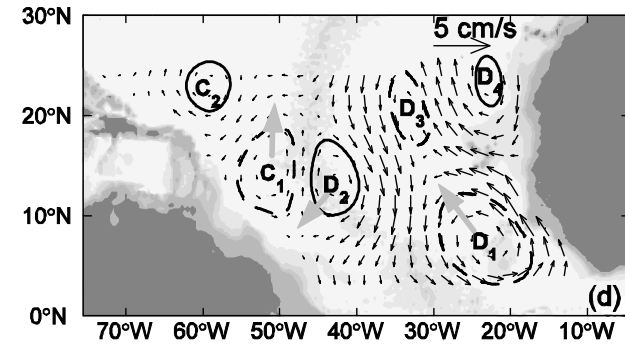
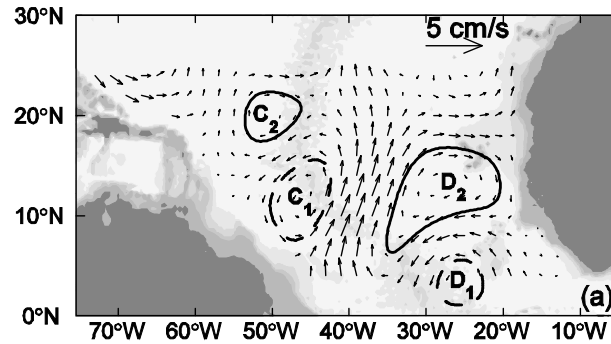


Correlation between wind Stress curl and streamfunction (solid: no-lag, dashed: 3 mon lag)



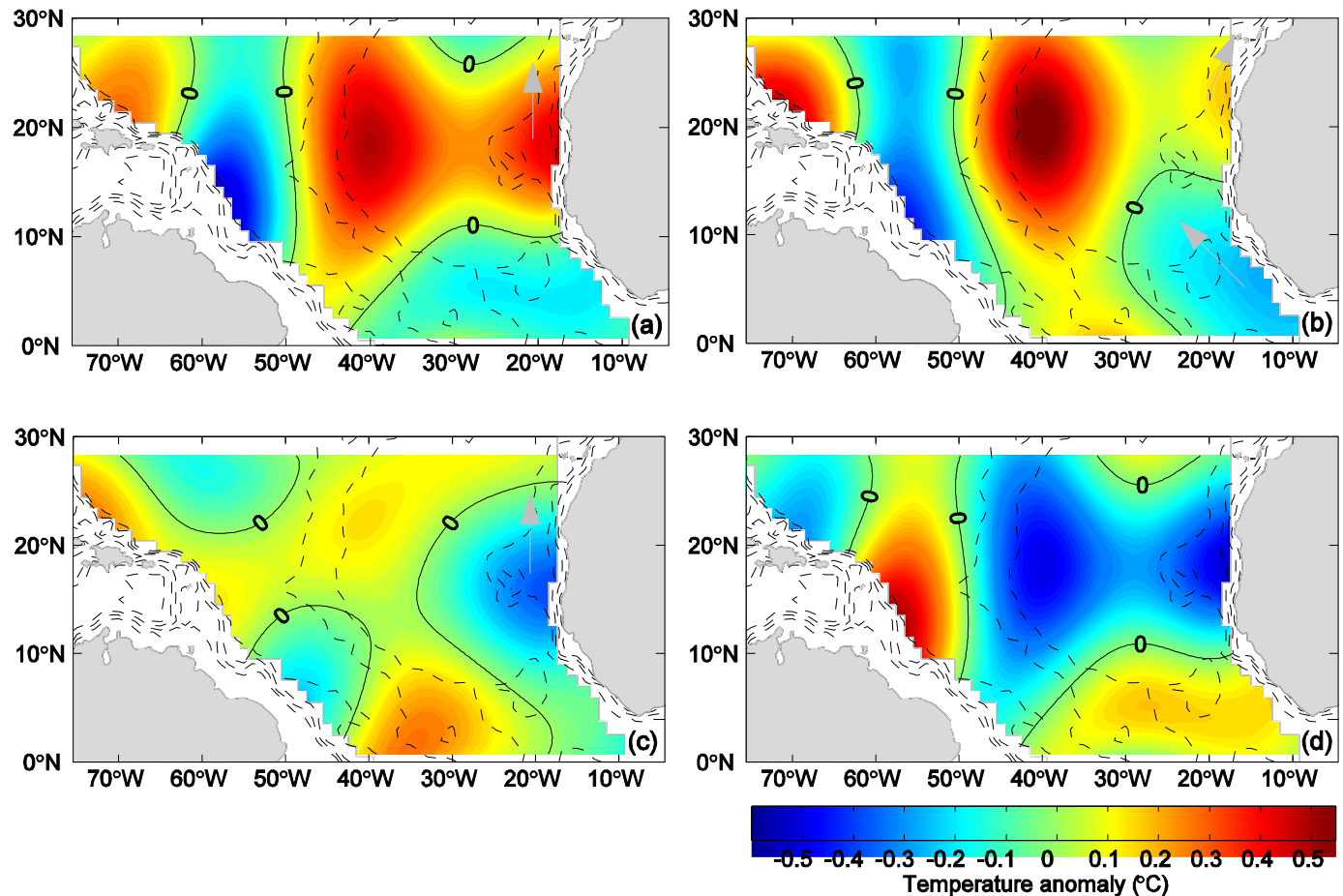
Semi-annual currents at 1000 m depth (2004)

(a) 5/15
(b) 5/30
(c) 6/14
(d) 6/29
(e) 7/13



Semi-annual monthly temperature anomaly at 950m depth

- a) Jun 04
- b) Aug 04
- c) Oct 04
- d) Dec 04.



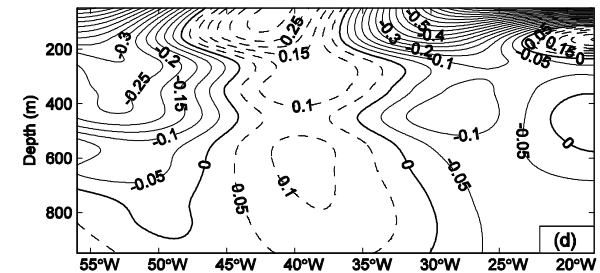
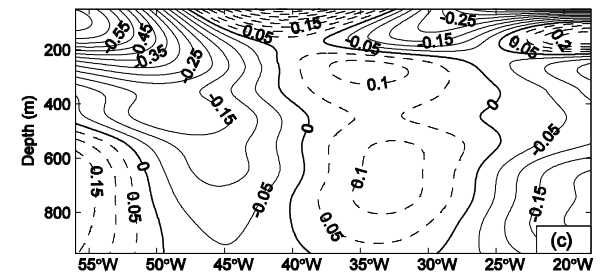
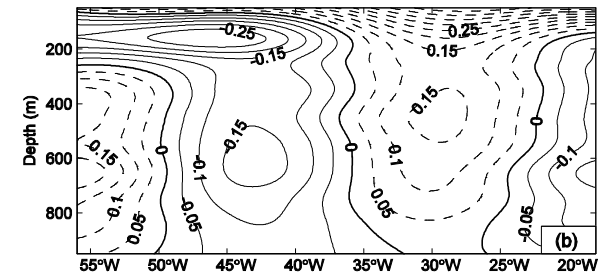
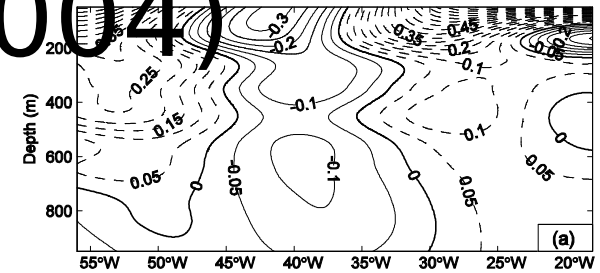
Semi-annual component of monthly temperature anomaly along 11°N (2004)

(a) 6/4

(b) 7/4

(c) 8/4

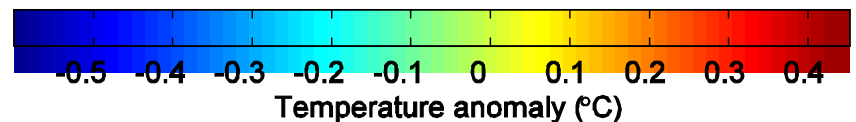
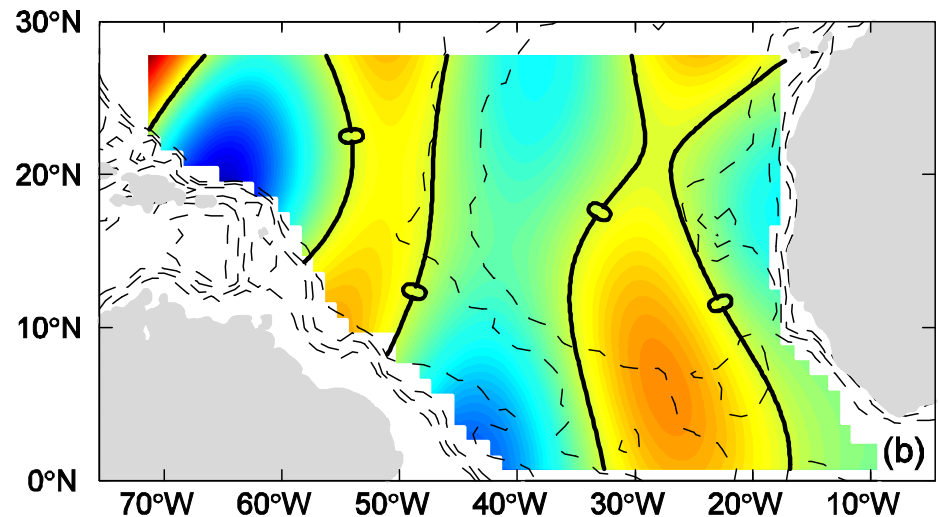
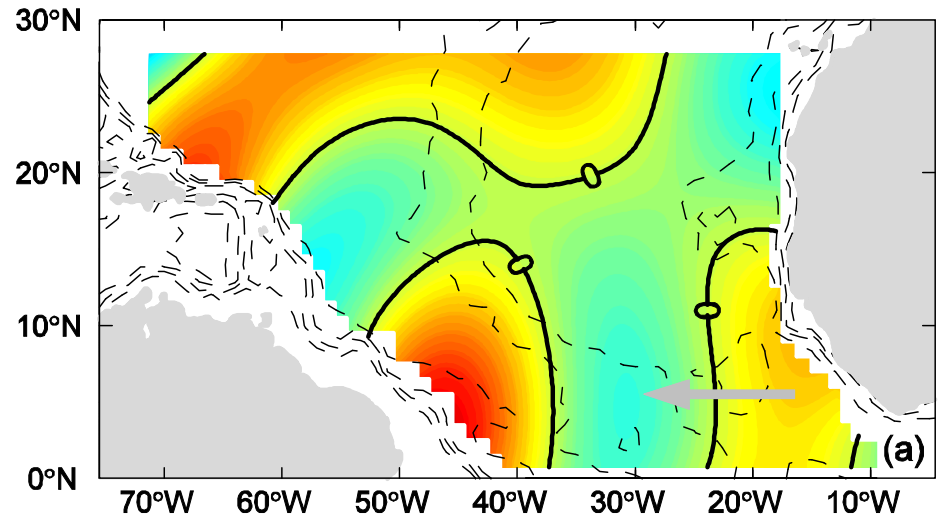
(d) 9/4



Semi-annual temperature anomaly at 550m depth (2004)

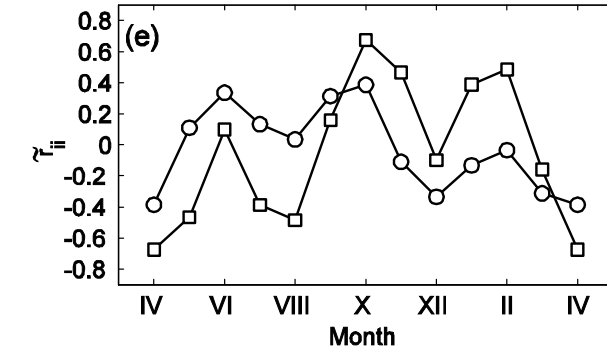
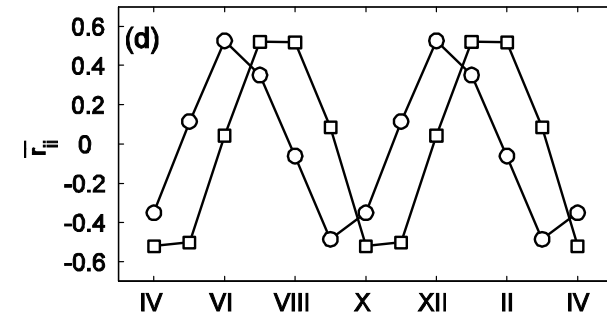
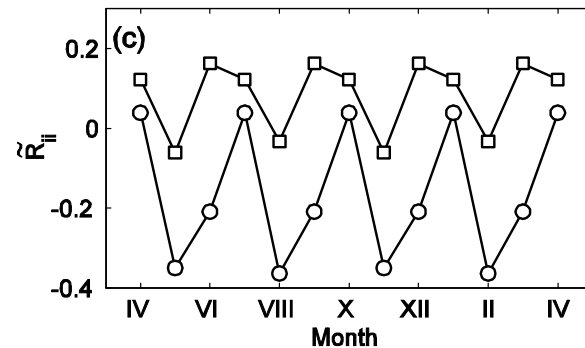
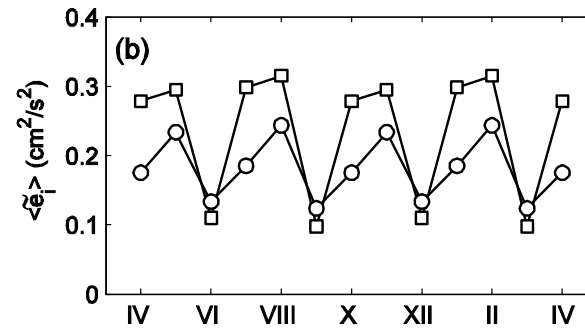
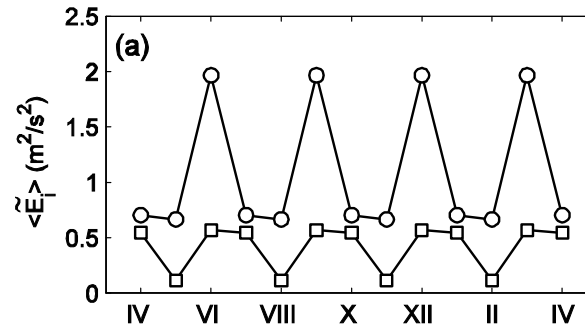
(a) 5/15

(b) 6/29



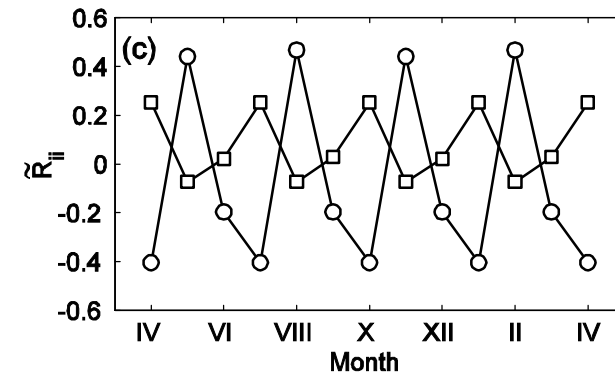
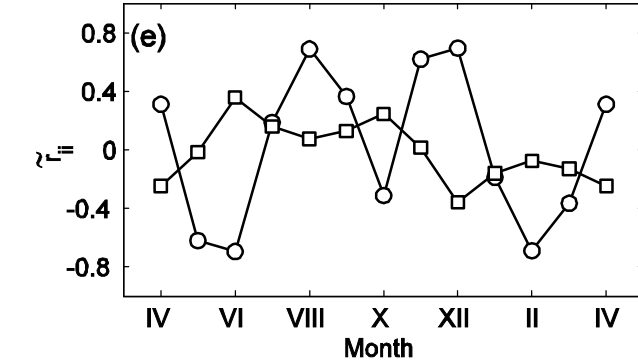
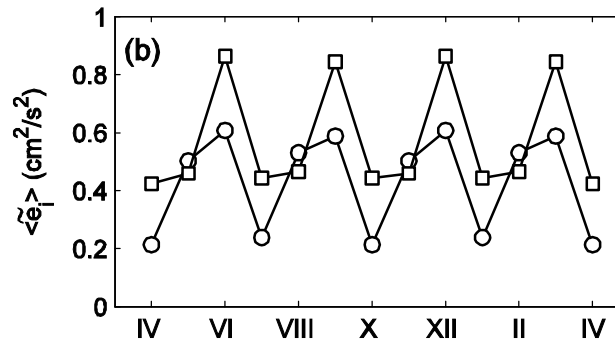
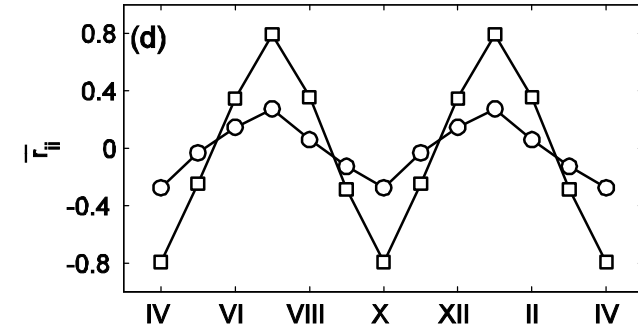
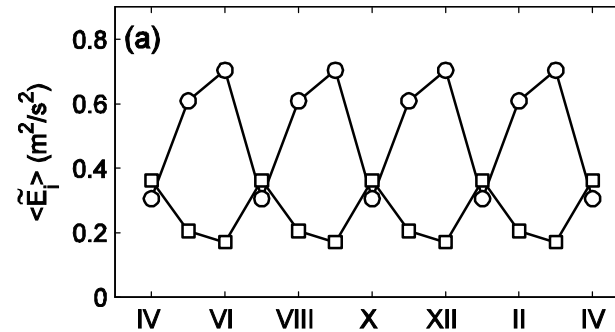
Semiannual Component in the Western Sub-Basin

- (a) wind KE
- (b) current KE
- (c) corr wind stress and currents
- (d) corr between semi-annual currents and mean wind
- (e) corr between semiannual currents and annual wind stress.



Semiannual Component in the Eastern Sub-Basin

- (a) wind KE
- (b) current KE
- (c) corr wind stress and currents
- (d) corr between semi-annual currents and mean wind
- (e) corr between semiannual currents and annual wind stress.



Results

- The annual and semi-annual unstable standing Rossby waves are detected in both the western and eastern sub-basins.
- The wind-driven Ekman pumping seems to be responsible for the standing wave generation in both the sub-basins.

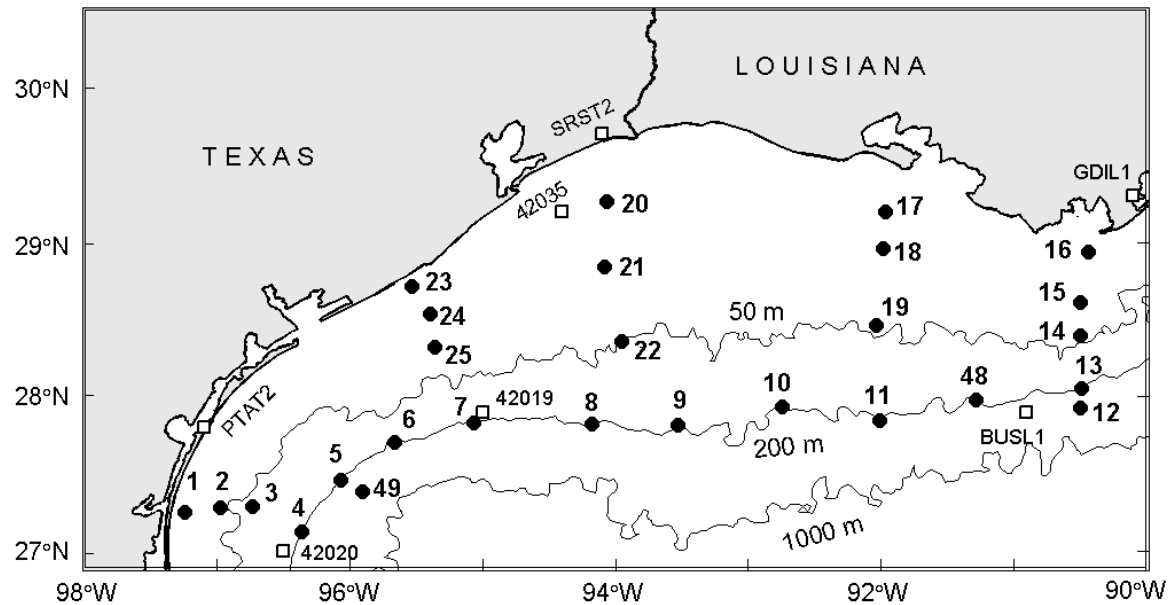
Example-3

OSD for Analyzing Combined Current Meter and Surface Drifting Buoy Data

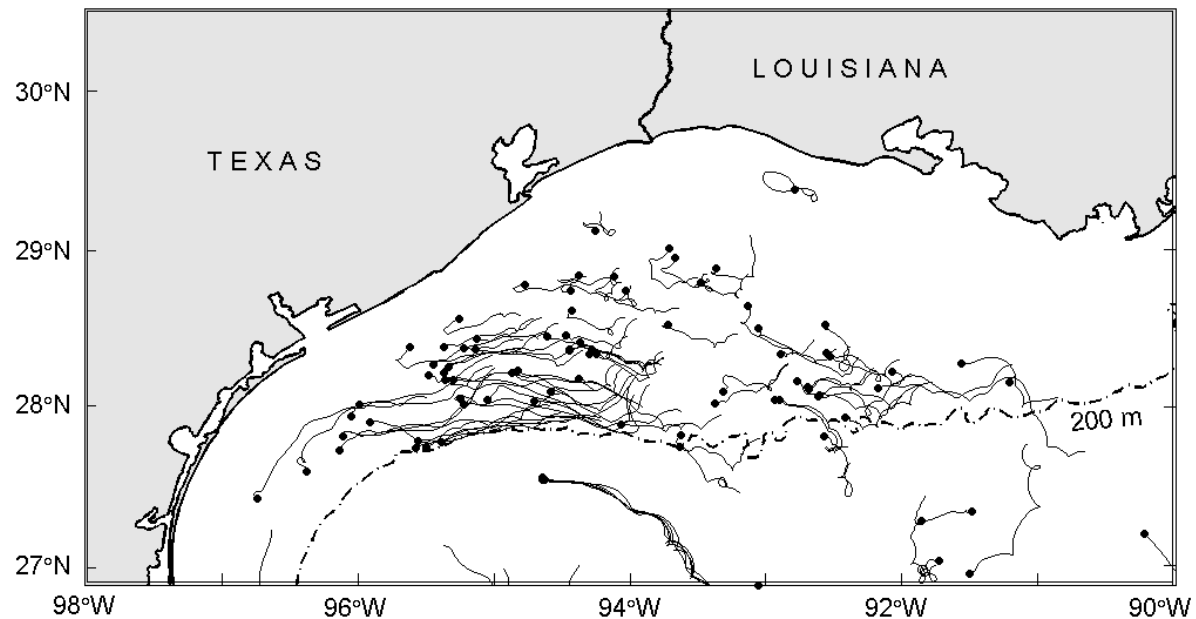
Ocean Velocity Observation

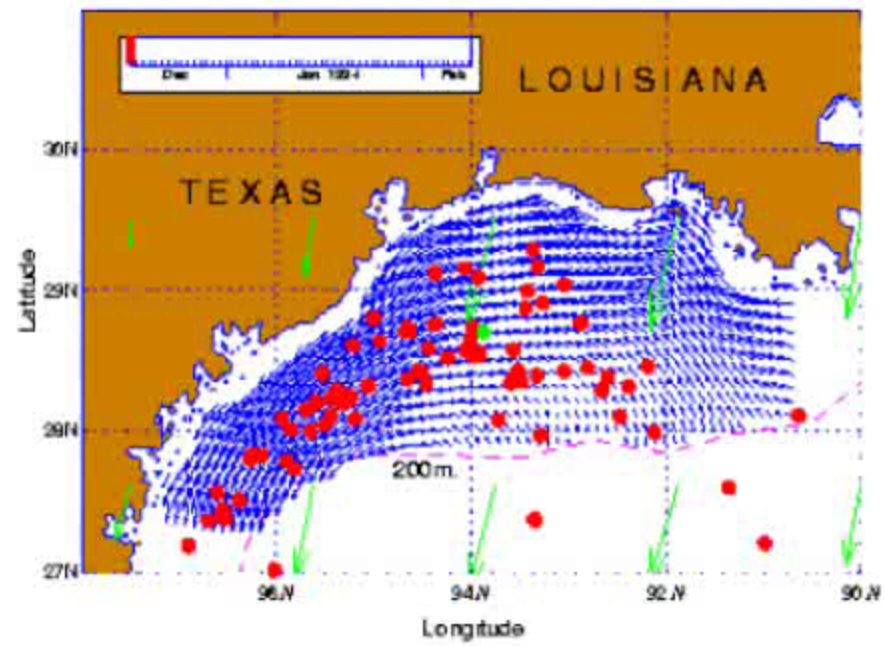
- 31 near-surface (10-14 m) current meter moorings during LATEX from April 1992 to November 1994
- Drifting buoys deployed at the first segment of the Surface Current and Lagrangian-drift Program (SCULP-I) from October 1993 to July 1994.

Moorings and Buoys

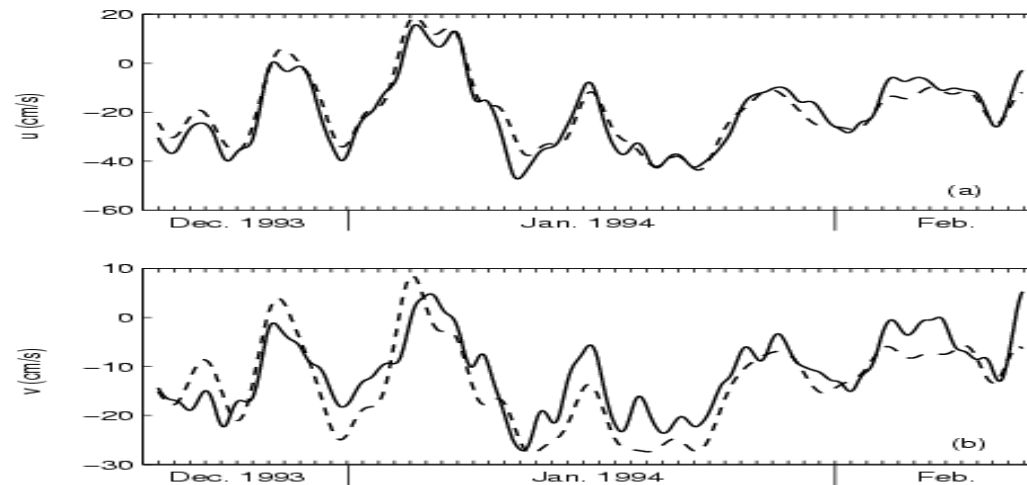


LTCS current reversal detected from SCULP-I drift trajectories.





Reconstructed and observed circulations at Station-24.



Probability of TLCS Current Reversal for Given Period (T)

- $n_0 \sim$ 0-current reversal
- $n_1 \sim$ 1-current reversal
- $n_2 \sim$ 2-current reversals
- $m \sim$ all realizations

$$P_0(T) = \frac{n_0}{m}, P_1(T) = \frac{n_1}{m}, P_2(T) = \frac{n_2}{m},$$

Fitting the Poisson Distribution

$$P_k(T) = \frac{1}{k!} (\mu T)^k \exp(-\mu T)$$

$$k=0, 1, 2$$

μ is the mean number of reversal for a single time interval

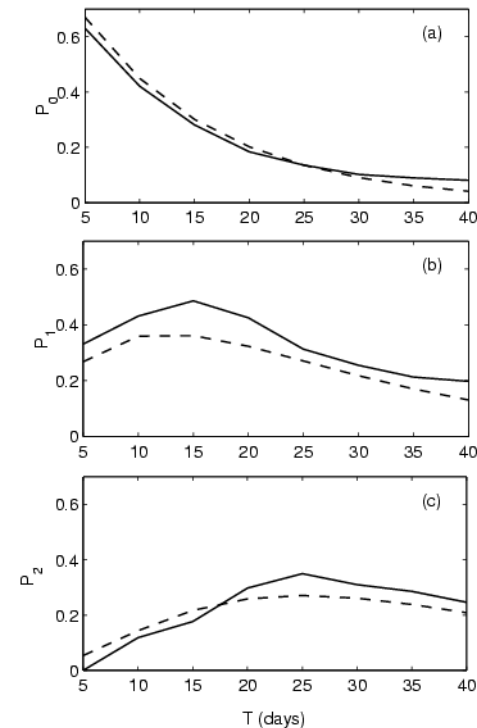
$$\mu \sim 0.08$$

Dependence of P_0 , P_1 , P_2 on T

For observational periods larger than 20 days, the probability for no current reversal is less than 0.2.

For 15 day observational period, the probability for 1-reversal reaches 0.5

Data – Solid Curve
Poisson Distribution Fitting –
Dashed Curve

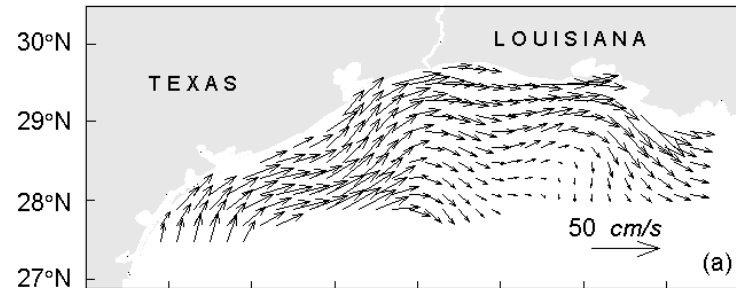


Time Interval between Successive Current Reversals (not a Rare Event)

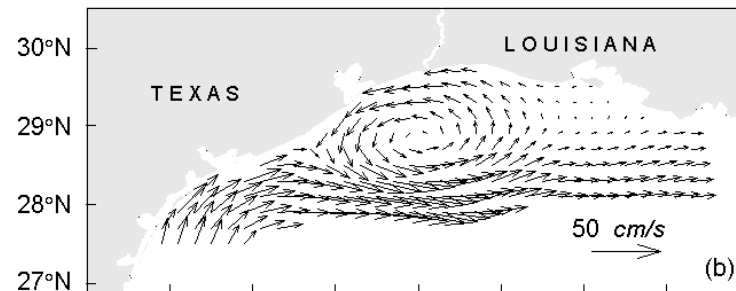
$$p(\tau) = \mu \exp(-\mu\tau)$$

LTCS current reversal detected from the reconstructed velocity data

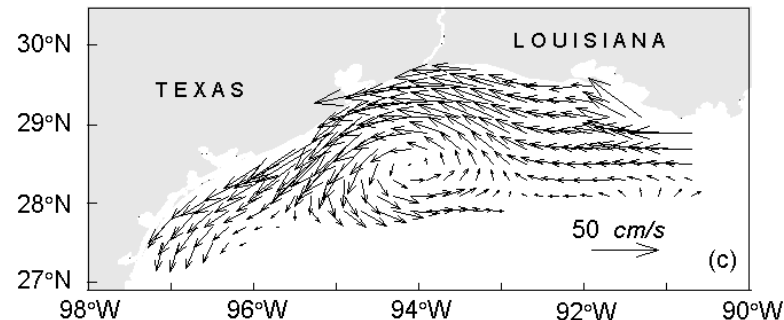
December 30, 1993



January 3, 1994



January 6,
1994



EOF Analysis of the Reconstructed Velocity Filed

EOF	Variance (%)		
	01/21/93-05/21/93	12/19/93-04/17/94	10/05/94-11/29/94
1	80.2	77.1	74.4
2	10.1	9.5	9.3
3	3.9	5.6	6.9
4	1.4	3.3	4.6
5	1.1	1.4	2.3
6	0.7	1.1	0.8

Mean and First EOF Mode

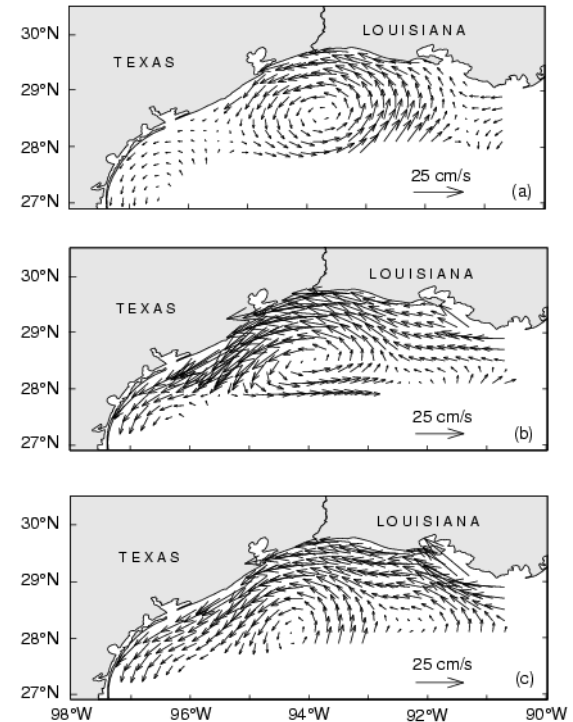
$$\tilde{\mathbf{u}}(x, y, t) = \bar{\mathbf{u}}(x, y) + A_1(t)\mathbf{u}_1(x, y),$$

Mean Circulation

1. First Period
(01/21-05/21/93)

2. Second Period
12/19/93-
04/17/94)

3. Third Period
(10/05-11/29/94)

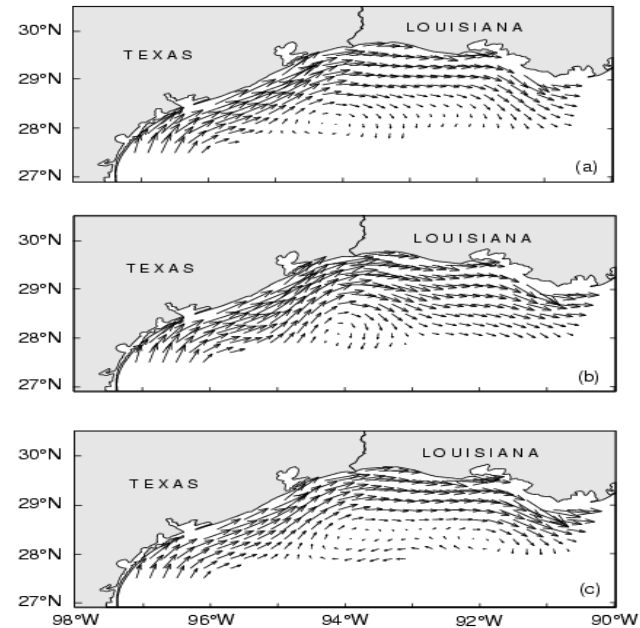


EOF1

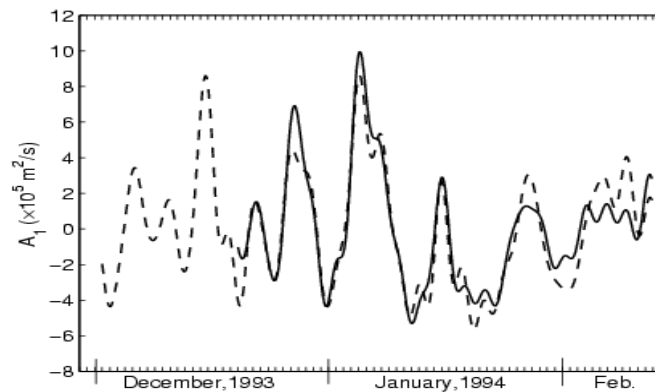
1. First Period
(01/21-05/21/93)

2. Second Period
(12/19/93-04/17/94)

3. Third Period
(10/05-11/29/94)



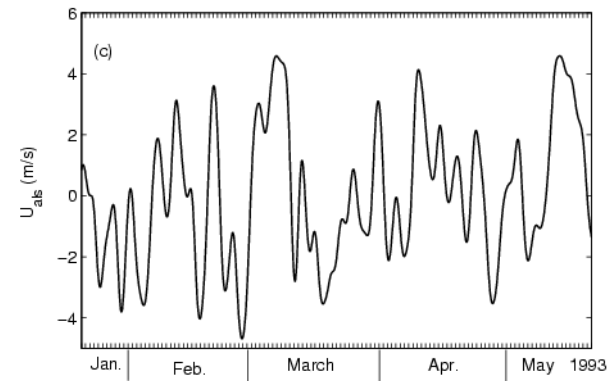
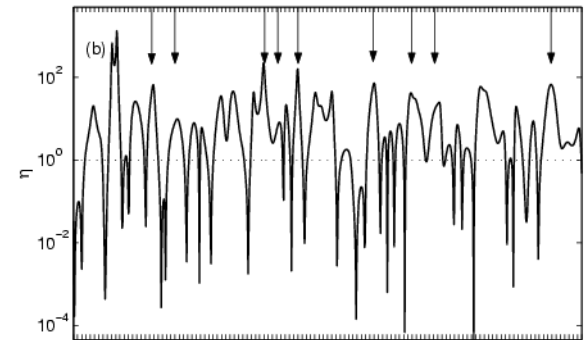
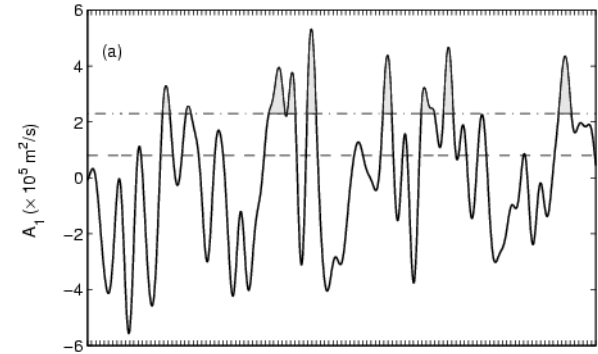
- Calculated $A_1(t)$
Using Current Meteorological
Mooring (solid)
and SCULP-1
Drifters (dashed)



- 8 total reversals observed

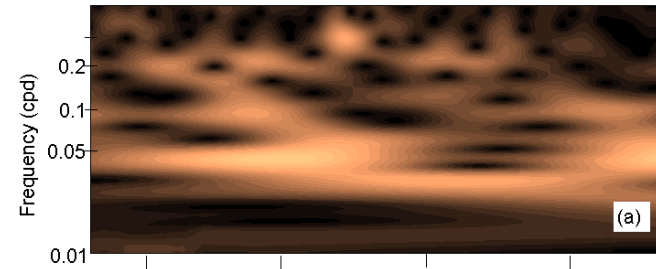
$$\eta = A_1^2 / \sum_{n=2}^6 A_n^2$$

- $U_{als} \sim$ alongshore wind

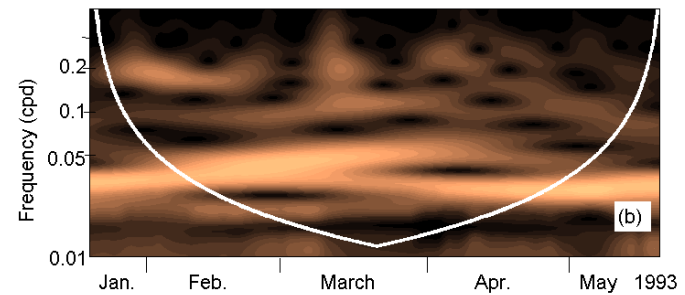


Morlet Wave

- $A_1(t)$



- U_{als}

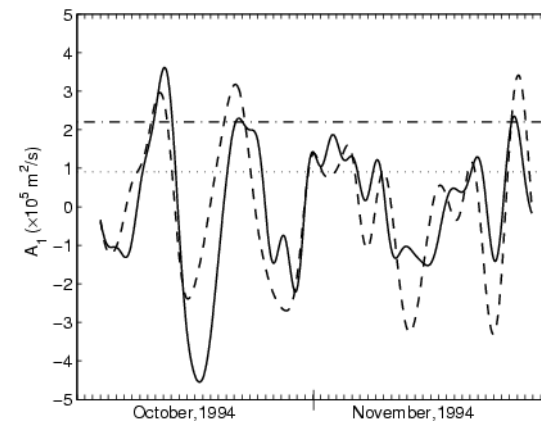


$$\Phi(t) = \pi^{-4} \exp(imt - t^2 / 2), \quad m = 6$$

Surface Wind Data

- **7 buoys of the National Data Buoy Center (NDBC) and industry (C-MAN) around LATEX area**

- Regression between
 - $A_1(t)$ and Surface
 - Winds
-
- Solid Curve
(reconstructed)
 - Dashed Curve
(predicted using
winds)



$$A_1(t) = \alpha[U(t) - \bar{U}] + \beta[V(t) - \bar{V}] + \gamma$$

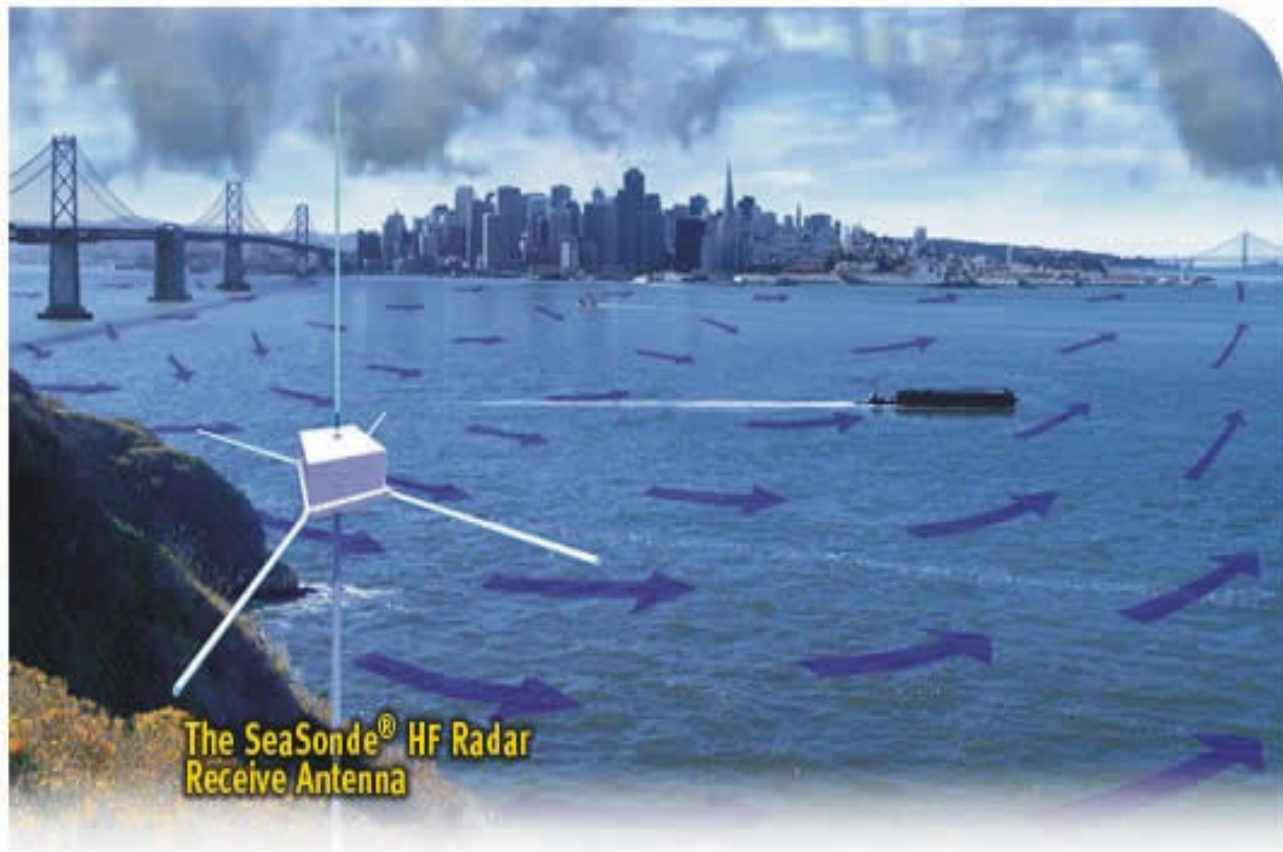
Results

- Alongshore wind forcing is the major factor causing the synoptic current reversal.
- Other factors, such as the Mississippi-Atchafalaya River discharge and offshore eddies of Loop Current origin, may affect the reversal threshold, but can not cause the synoptic current reversal.

Part-4

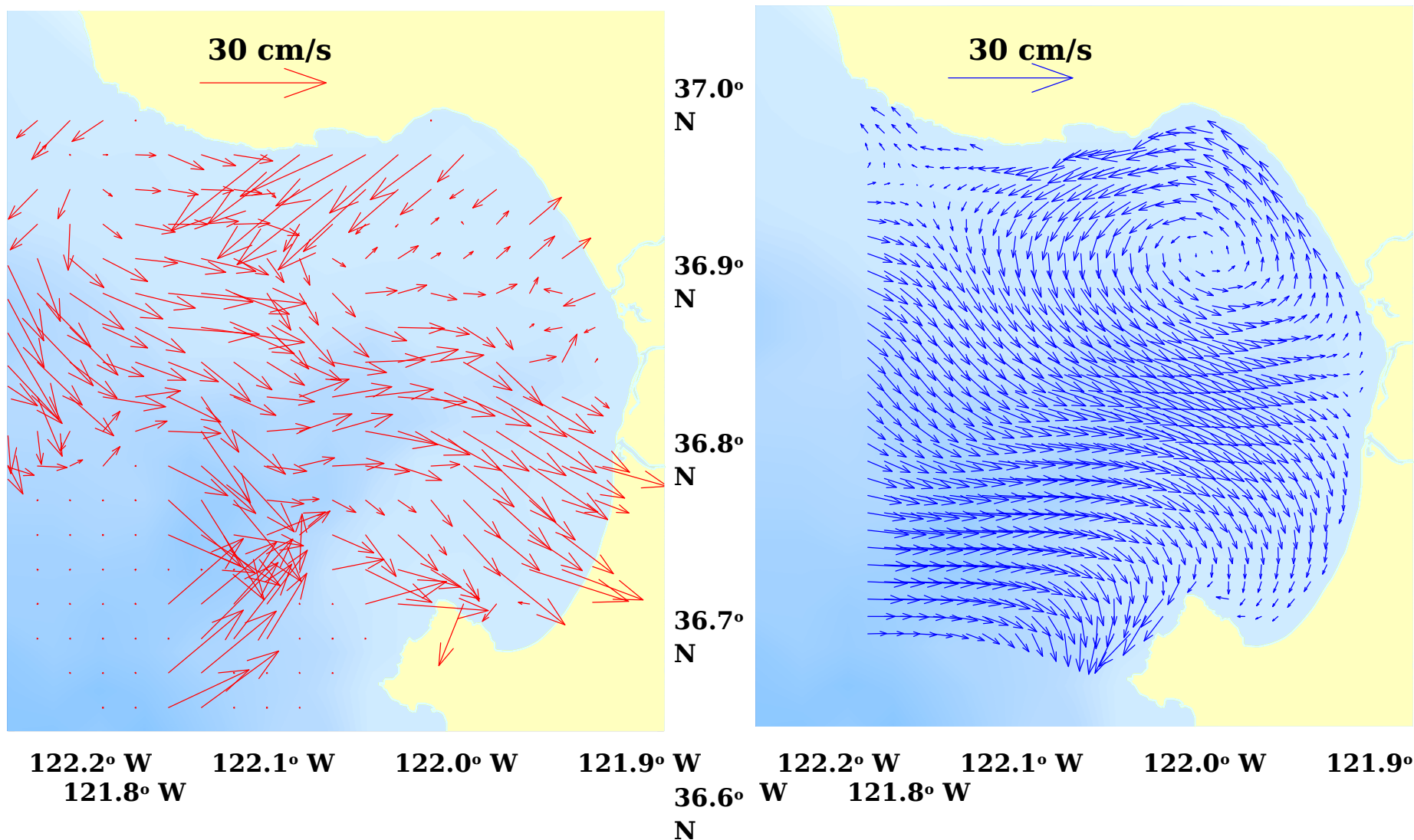
OSD for Analyzing CODAR Data

CODAR



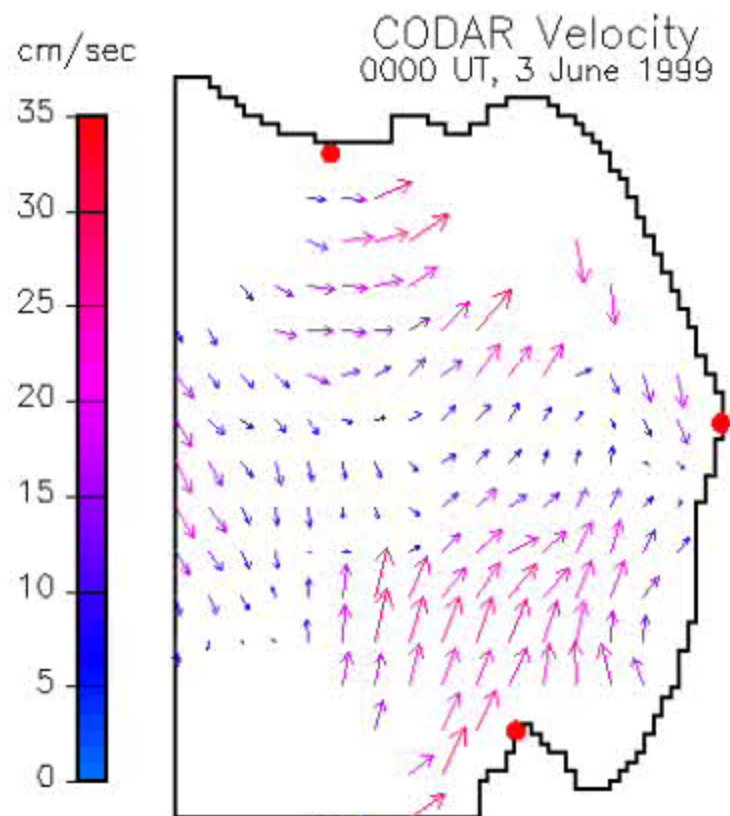
Monterey Bay



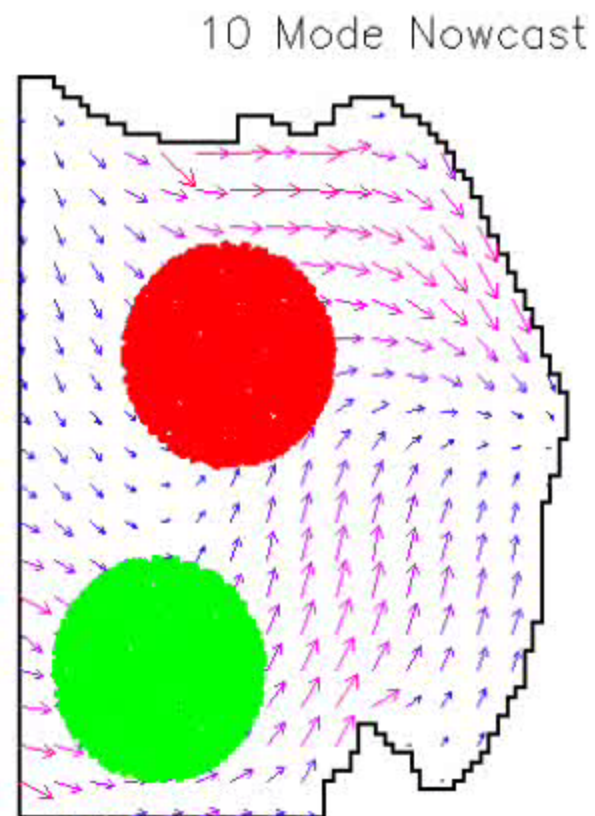


Place for comments: left - radar derived currents for 17:00 UT December 1, 1999

right - reconstructed velocity field.



→ 25 cm s⁻¹



Conclusions

- OSD is a useful tool for processing real-time velocity data with short duration and limited-area sampling especially the ARGO data.
- OSD has wide application in ocean data assimilation.